

PASSING THE TEST: A MODEL-BASED ANALYSIS OF SAFE SCHOOL-REOPENING STRATEGIES

Alyssa Bilinski MS¹ · Joshua A. Salomon, PhD² · John Giardina¹
Andrea Ciaranello MD, MPH³ · Meagan C. Fitzpatrick, PhD⁴

Abstract

Background. The COVID-19 pandemic has induced historic educational disruptions. In December 2020, at least two-thirds of US public school students were not attending full-time in-person education. The Biden Administration has expressed that reopening schools is a priority.

Objective. To compare risks of SARS-COV-2 transmission in schools across different school-based prevention strategies and levels of community transmission.

Design. We developed an agent-based network model to simulate transmission in elementary and high school communities, including home, school, and inter-household interactions.

Setting. We parameterized school structure based on average US classrooms, with elementary schools of 638 students and high schools of 1,451 students. We varied daily community incidence from 1 to 100 cases per 100,000 population.

Patients (or Participants). We simulated students, faculty/staff, and adult household members.

Interventions. We evaluated isolation of symptomatic individuals, quarantine of an infected individual's contacts, reduced class sizes, alternative schedules, staff vaccination, and weekly asymptomatic screening.

Measurements. We projected transmission among students, staff and families during one month following introduction of a single infection into a school. We also calculated the number of infections expected for a typical 8-week quarter, contingent on community incidence rate.

Results. School transmission risk varies according to student age and community incidence and is substantially reduced with effective, consistent mitigation measures. Nevertheless, when transmission occurs, it may be difficult to detect without regular, frequent testing due to the subclinical nature of most infections in children. Teacher vaccination can reduce transmission to staff, while asymptomatic screening both improves understanding of local circumstances and reduces transmission, facilitating five-day schedules at full classroom capacity.

Limitations. There is uncertainty about susceptibility and infectiousness of children and low precision regarding the effectiveness of specific prevention measures, particularly with emergence of new variants.

Conclusion. With controlled community transmission and moderate school-based prevention measures, elementary schools can open with few in-school transmissions, while high schools require more intensive mitigation. Asymptomatic screening can both reduce transmission and provide useful information for decision-makers.

¹Interfaculty Initiative in Health Policy, Harvard Graduate School of Arts and Sciences (abilinski@g.harvard.edu)

²Center for Health Policy & Center for Primary Care and Outcomes Research, Stanford University School of Medicine

³Medical Practice Evaluation Center and Division of Infectious Disease, Massachusetts General Hospital, Harvard Medical School

⁴Center for Vaccine Development and Global Health, University of Maryland School of Medicine (meagan.fitzpatrick@som.umaryland.edu)

Acknowledgements. The authors were supported by the Centers for Disease Control and Prevention through the Council of State and Territorial Epidemiologists (NU38OT000297-02; AB, JAS), the National Institute of Allergy and Infectious Diseases (R37AI058736-16S1; ALC and K01AI141576; MCF), the National Institute on Drug Abuse (3R37DA01561217S1; JAS). The papers' contents are solely the responsibility of the authors and do not necessarily represent the official views of the funders. Model code is available at <https://github.com/abilinski/BackToSchool2><https://github.com/abilinski/BackToSchool2>.

INTRODUCTION

COVID-19 prompted an unprecedented number of school closures during the spring of 2020. All 50 states recommended or mandated public school closures, impacting at least 124,000 US schools and over 55.1 million students (1). Reopenings have proven challenging, and as of December 9, 2020, only 33% of K-12 students were offered full-time in-person learning, a figure that has since the fall (2). Where in-person schooling has been offered, a substantial percentage of families have opted out, including more than half in parts of Orange County, Florida and 70% in New York City (3,4). Many parents and advocates have objected strongly to school closures (5), and the Biden administration has stated a priority to rapidly facilitate safe, in-person school reopening in a manner that gains the trust of students, parents, and teachers (6).

Debates around school reopening have been heated and often invoked seemingly contradictory evidence about its safety. For example, a number of well-studied cases in school settings have found minimal secondary transmission (7–9). Nevertheless, significant school clusters have also been documented, particularly in Israel and in parts of the United States (10–12), and some observational studies have suggested that school closures may have substantially reduced transmission (13). Proponents of reopening schools point to evidence that COVID-19 is most often mild in children (15), while others have expressed concern that infection may lead to severe outcomes in some children, as well as among staff, families, and the community. The discord plays out similarly on a macro scale: a number of Asian and European countries reopened schools with physical distancing when community transmission was low and reported negligible increases in transmission (16,17). Some, including France, the United Kingdom, and Ireland, kept schools open during the fall wave and have reversed surging transmission by closing other sectors, although the latter two closed schools following the emergence of variant B.1.1.7. (18,19). Others, such as Austria, the Czech Republic, and South Korea, have closed schools to address rising case burdens (18).

Nevertheless, there is little debate that the benefits of in-person education are substantial, particularly amidst reports of high levels of remote absenteeism, increased depression, anxiety, and suicidality, and parent concerns around educational quality (20–22). Beyond educational and mental health outcomes, opening schools can also improve access to social services for children and labor market outcomes for working parents, especially women (23–27). In order to do so safely, it is critical to take a comprehensive account of

the full array of sometimes contradictory evidence and identify a robust set of policies to mitigate risks.

We simulated SARS-CoV-2 transmission dynamics in elementary and high schools, to characterize how school transmission may occur either as isolated or sustained outbreaks. While several recent papers have discussed the impact of school reopening on the broader epidemic under different mitigation strategies (28–30), our work focused on the risk that transmission will occur on a school campus and spread to households of students and educators/staff. We emphasize uncertainty in transmission risk (rather than focusing on the average) as well as the observability of infections to school and public health staff, allowing us to reconcile apparent contradictions in the evidence. We evaluated outcomes under varying combinations of community incidence, in-classroom mitigation efforts, testing practices, and staff vaccination. We project that consistently implemented mitigation, including masking and distancing, can drastically reduce the risk of school-based transmission, particularly in elementary schools. Nevertheless, high rates of subclinical disease, especially in children and adolescents, may render transmission difficult to connect to schools via contact tracing, and imperfect implementation of mitigation strategies or stochastic error (“bad luck”) can produce outbreaks even in well-managed systems. We find that teacher vaccination and asymptomatic screening can help support safe return to 5-day schooling for all children with only targeted closures, even amidst moderate ongoing community transmission.

METHODS

We developed an age-specific, agent-based network model of COVID-19 transmission in elementary and high school communities (Figure 1). We incorporated interactions within schools and homes, as well as those between households in childcare or social interactions outside of the school setting. Because data are particularly inconclusive on transmission among middle school-aged children, we did not model middle schools explicitly; accumulating evidence suggests that they may be more similar to high schools than elementary schools with respect to susceptibility and infectiousness (31,32).

SCENARIOS

We simulated different combinations of interventions, incorporating strategies that target the three axes of general infection control in schools; COVID-19-specific countermeasures; and scheduling/cohorting with parameters detailed in Table 1.

General infection control in schools:

1. **Low uptake:** Schools implemented no or minimal general infection control measures, such as masking and distancing.
2. **Medium uptake:** Schools implemented masking and distancing, and adherence was moderate such

that the risk transmission given infectious contact was $\frac{2}{3}$ that experienced in a school with low uptake.

3. **High uptake:** Schools implemented masking and distancing and adherence was high, such that the risk of transmission given infectious contact was $\frac{1}{3}$ that experienced in a school with low uptake.

COVID-19-specific countermeasures:

1. **Symptom-based self-isolation (“Symptomatic isolation”):** Individuals in the school were screened for symptoms daily, and those who developed clinically-recognizable symptoms did not attend school.
2. **Diagnostic testing plus classroom notification (“Classroom quarantine”):** Individuals who developed symptoms were isolated and immediately tested. The school received results within a day, and all classroom(s) associated with an individual who tested positive were notified and closed for 10 days, per updated/recent CDC guidelines (32).
3. **Teacher vaccination with an infection-blocking vaccine (“Staff vaccination”):** In addition to classroom quarantine, teacher susceptibility was reduced to 33% of baseline, roughly assuming that 75% of teachers receive an infection-blocking vaccine with 90% effectiveness (33–35).
4. **Weekly asymptomatic screening (“Weekly screening”):** In addition to classroom quarantine, asymptomatic students and/or staff at each school were tested weekly. We assumed test acceptance was 90%, sensitivity was 90% during the infectious period, and results were available within 24 hours. Upon receipt of a positive result, infected individuals isolated outside of school for 10 days based on CDC guidelines, and siblings and classroom members of an individual who tests positive were notified and quarantined for 10 days.

Scheduling/Cohorting:

1. **5-day schedule:** This scenario simulated a traditional 5-day, in-person learning schedule, allowing for classroom contacts, staff-staff interactions (10 per day), and random contacts between school members (20 per day). Related arts and special education teachers had revolving contact with 5 classrooms per day.
2. **Cohorting:** This scenario again assumed 5 days of in-person learning for all students, but with restricted out-of-classroom contacts, including separation of classes for lunch and recess. Students and primary teachers continued to have sustained classroom contacts. We assumed a 50% decrease in the number out-of-classroom contacts during the school day and that related arts were taught remotely.
3. **Half class sizes:** All students attended school five days per week, but in classes of half their typical size. To accommodate this, the number of teachers was doubled. Limited contacts as in (2) were also maintained.
4. **Hybrid scheduling (A/B):** Classes were subdivided into 2 cohorts, and students attended school for

2 days per week, either Monday/Tuesday or Thursday/Friday. Elementary school children in the same household attending the same school were sorted into the same cohort. All staff were physically present each instructional day. In sensitivity analysis, we considered alternative hybrid schedules. Limited contacts as in (2) were also maintained.

MODEL STRUCTURE

Households

We generated a set of synthetic households from US population data (36) (Supplement). For elementary schools, we sampled from households containing at least one child aged 5 to 10 to identify siblings attending the same school. For high schools, we sampled from households with students aged 14 to 17. For each student, we included two adults in the household, based on the average number of household members over 25. For each staff member, we also included a household adult contact, representing a partner or roommate with whom they had close contact.

Elementary schools

To simulate a representative school, we identified the average school size weighted by student population (37). We simulated an average-sized elementary school of 638 pupils from 500 households, of which 114 households had more than one child in the school. We set five classes per grade level, an average class size of 21 children, and one primary teacher per class. We also incorporated 30 additional staff, to reflect roles such as administrators, counselors, cafeteria staff, custodians, special education teachers, and teachers of related arts (or ‘specials’, e.g. music and art). These adults were assigned either in-classroom roles ($n = 15$, including special education and related arts teachers who are in rotating contact with entire classes of students) or out-of-classroom roles ($n = 15$, interacting with fewer individuals outside the classroom setting).

High schools

We simulated an average-sized high school of 1451 students in 1225 households, with 142 households having more than one student in the school (37). Students rotated among 8 class periods per day (23 students and 1 teacher), with the distribution of students to class periods chosen randomly within grade levels but repeated daily for the full simulation. The high school had 45 additional staff with out-of-classroom roles and 15 with in-classroom roles. In both elementary and high schools, teachers were modeled to have additional staff contacts per day (e.g., contacts in break rooms or offices).

Out-of-school interactions

Reflecting social interactions and out-of-school childcare, our base case analysis assumed that each family interacted with 1 additional family on each day they did not attend school, randomly reassorted each day.

We varied this number in sensitivity analysis from 0 to 9 families per day out of school. We assume that when families mix, no more than 2 adults attend at a given time, who are randomly selected.

TRANSMISSION AND EPIDEMIOLOGICAL PARAMETERS

At each daily time-step, we modeled dyadic interactions between individuals according to household, classroom, school, and childcare relationships, drawing parameter values from the distributions specified in Table 1. A SARS-CoV-2-infected individual i transmitted to susceptible individual j at time t with Bernoulli probability equal to:

$$p_{ijt} = c_{ijk} q_k s_j r_i a_i d_i$$

where c_{ijk} was an indicator variable equal to 1 if individuals i and j had contact type k at time t , q_k was the probability of transmission given one day of contact type k , r_i was the relative infectiousness of individual i (compared to full adult infectiousness), s_j was the relative susceptibility of individual j (compared to full adult infectiousness), and a_i indicated whether individual i had asymptomatic disease and d_i was a dispersion factor representing individual-level heterogeneity in transmissibility.

Secondary attack rate (SAR) (q_k)

Secondary attack rates (SARs) for SARS-CoV-2, the probability that a person with SARS-CoV-2 transmitted it to an individual they contacted, vary by contact type: household, classroom, random school, and out-of-school social/childcare contacts. Household SARs have been most frequently estimated in a number of contact tracing studies. While there is substantial variation across geographic locations, likely corresponding to cultural norms and precautions adopted, we assumed a 20% household adult-adult secondary attack rate over the full duration of an infection, translating to approximately 4% per day (38,39). This assumed minimal in-home quarantine measures but less household mixing than during full shelter-in-place orders.

We used the SAR from household contact tracing studies as a rough proxy for the per-day adult-adult attack rate in non-household settings, following close sustained contact in the absence of interventions such as masks and distancing. For school based transmission, we allowed adult-adult attack rates to reach up to 3% per day for scenarios with low uptake of masks and distancing, a downward adjustment from the household attack rate to account for less close contact in professional settings. We developed the scale of reductions in transmission from mitigation measures in line with observations from in household settings with high prevention measures (31,40) and based on effectiveness analyses for measures like masks (41,42).

Relative susceptibility (s_i) and infectiousness in children (r_i)

Household contact tracing studies suggest that adults are likely more susceptible to COVID-19 and more apt to transmit it when infected than children, although data, particularly for the latter, are equivocal and limited by availability and timing of testing (43). Several sources indicate that any differences wane by the teenage years, possibly as early as age 10 (8,10,31,44,45). We assumed that elementary school children were half as susceptible as adults while high school children were equally susceptible. Data are similarly limited to inform transmission from children with COVID-19. We further specified that elementary school children are half as infectious as adults and high school students were equally as infectious as adults (31,44). We varied these assumptions in sensitivity analyses. We discuss the data underlying this assumptions as well as comparisons to other modeling work and documented in-school transmissions in the Supplement.

Asymptomatic transmission (a_i) and overdispersion (d_i)

We assumed that individuals with fully asymptomatic disease transmit COVID-19 at half the rate of those with any symptoms (46). While reported heterogeneity in transmission may be partially driven by differences in contact rates, we sample individual transmissibility according to a lognormal distribution to allow for some variation by individual characteristics (e.g. viral load) in adults who have been more widely identified as potential superspreaders (47,48).

MODEL IMPLEMENTATION

We first evaluated the downstream impact of a single infectious individual. We randomly designated one member of the school community as infected, starting on a random day of the week. We assessed the spread of the virus over 30 days, as in most cases either all infections were resolved over that time horizon or the spread was sufficiently large that additional public health measures (e.g. school closures) would be expected to be adopted during that time. For each scenario, we ran our model 2000 times and summarized:

1. Mean number of infections generated in the school over 30 days after index case
2. Percentage of scenarios without transmission from the index case
3. Percentage of scenarios with more than 5 in-school transmissions

We also describe the composition of secondary cases (proportion occurring in students, staff, and family members of students or staff).

Second, we quantified SARS-CoV-2 infections among our school population across a typical school quarter, given a constant community incidence. On each day over 8 weeks, every susceptible individual has a probability of becoming spontaneously infected outside of school that is equivalent to an age-adjusted

community per capita daily incidence, distinct from their contact-dependent risk within the school network. When schools are remote, we assumed, similar to the analysis of hybrid schools, that each family interacted with 1 additional family on each day they did not attend school. For each scenario, we summarized:

1. Cumulative incidence overall and by type
2. Incremental incidence compared to remote learning

Transmission control threshold: We defined a reopening strategy as controlling transmission in a group (students, educators/staff, or families) if it resulted in less than a one point increase in the percentage of the group which became infected, compared to remote learning. This threshold is consistent with thresholds used by similar papers and with the generally consensus objective of minimizing in-school transmission (30,49).

RESULTS

IN-SCHOOL TRANSMISSION AFTER THE INTRODUCTION OF A SINGLE INFECTION

In-school mitigation

In elementary schools with low mitigation and symptomatic isolation under a 5-day schedule, we projected an average of 2 secondary cases over 30 days following infection of a single index case (Figure 2A, left panel). This fell to 1 case with medium mitigation, and further to 0.4 cases with high mitigation. Under symptomatic isolation, transmission was most reduced by replacing the 5-day schedule with an A/B schedule (to 0.1-0.4 secondary cases depending on uptake of in-school prevention). With high mitigation, there were on average <0.5 secondary transmissions per case in all scenarios.

In high schools, where we specified that students acquired and transmitted infection at the same rate as adults, we found the potential for larger outbreaks after a single introduction into the school, particularly when uptake of in-school prevention was low (Figure 2B). For example, with low mitigation and classroom quarantine under a 5-day schedule, we projected 22 secondary cases in the school community over a 30-day period (in the absence of public health responses that would likely ensue before 30 days, such as school closure). High uptake of in-school mitigation reduced this markedly, to 1.8 cases.

QUARANTINE, TEACHER VACCINATION, AND SCREENING

In elementary schools, where children were assumed to be less susceptible and infectious, classroom quarantine had a modest impact on transmission, although its impact was greater if the index case was a teacher (Figure S1). In high schools, classroom quarantine reduced projected average transmissions

by a factor of 0.23 under low mitigation, but had smaller impact with high mitigation. We found a low impact of teacher vaccination on overall transmission, but substantial impact on transmission to teachers. In elementary schools, teacher vaccination reduced secondary infections over 30 days to 95% overall and 96% of the average without vaccination, in elementary and high schools respectively. In both settings, it reduced staff secondary incidence to a third of its initial rate, although the reduction was slightly greater if the index case was a teacher.

Weekly screening was projected to reduce secondary cases by a large degree compared to symptomatic isolation, classroom quarantine, or teacher vaccination: to an average of 0.2-0.8 with a 5-day schedule in elementary schools (varying by uptake of in-school prevention) and to 1.0-4.5 in high schools (Figure 2A, right panel). The impact of weekly screening following a single introduced case was greatest for settings with low mitigation: weekly screening reduced average projected secondary cases from 2.0 to 0.8 in elementary schools and from 80 to 4.5 in high schools.

DETECTION OF SCHOOL-RELATED TRANSMISSION

Across all scenarios, for elementary schools, we estimated that 73% of secondary cases would occur in students, 19% in families, and 8% in teachers and staff. For high schools, 77% would occur in students, 19% in families, and 4% in teachers and staff (Figures S2 and S3). Although staff are the smallest fraction of these groups, they had the highest per capita incidence following introduction of an infection. As children are less likely than adults to have symptoms, we projected that 14% of all secondary infections in elementary school communities and 15% in high school communities would be clinically symptomatic and therefore detectable without asymptomatic testing (Figure S4). With a more expansive definition of any symptoms, these percentages increased to 28% and 29% respectively. We discuss comparisons to documented in-school transmissions in the Supplement.

STOCHASTIC VARIATION IN SECONDARY TRANSMISSION

Re-running the simulation 2000 times with each set of parameters, we observed considerable variability across possible outcomes (Figure 3). In elementary schools with high mitigation and symptomatic isolation under a 5-day schedule, there was a 79% probability of zero secondary cases over a 30-day period (Figure 3A, left panel). However, there was a 0.3% probability of five or more secondary cases. The chance of >5 secondary cases occurring was higher when mitigation uptake was lower (Figure 3A, left and middle panels), reaching a 50% probability of zero secondary cases, and an 11% probability of five or more under low mitigation. In high schools with symptomatic isolation, the probability of no secondary cases similarly ranged from 19-52%, and the probability of 5 or more ranged from 13-71%, with long tails of outliers.

However, with any interventions, this long tail of large outbreaks was substantially reduced.

TRANSMISSIONS OVER THE COURSE OF THE SEMESTER

With any of the modeled scenarios (symptomatic isolation, classroom quarantine, teacher vaccination, and weekly screening), both 5-day and A/B schedules increased transmission compared to remote learning. This increase was greater when community incidence was higher, especially considering infection risk among staff (Figures 4-5).

Assuming that additional measures such as school closure were not taken within the 30 days, Figures 4 and 5 display the values of community incidence at which our predefined threshold for “controlling transmission” was met. In elementary schools with medium mitigation under a 5-day schedule, all strategies met this threshold among when community incidence fell below 25 cases per 100,000 per day. The control threshold for the full population was exceeded at 50/100,000/day with symptomatic isolation (cumulative incidence 7.8%, increment 1.002%) and at 100/100,000/day with classroom quarantine (total incidence 15%, increment 1.03% for classroom quarantine and 1.7% for symptomatic isolation). Evaluating the threshold specifically among teachers, the control threshold was never met when community incidence was ≥ 25 per 100,000/day, unless teachers were vaccinated (total incidence: 6.2%, increment 1.9% with classroom quarantine). With both high mitigation and teacher vaccination, strategies met the control threshold among teachers at all rates up to 100 per 100,000 cases/day.

In high schools, stronger mitigation or prevention strategies would be required to meet the control threshold. Under medium mitigation and community incidence of 100 cases per 100,000/day, only weekly screening met this threshold for the full population. Under high mitigation, all strategies except symptomatic isolation met this threshold. For the teacher control threshold in high schools, high mitigation combined with either weekly screening or teacher vaccination kept the increase in teacher cumulative incidence below the control threshold at 50 cases per 100,000/day.

SENSITIVITY ANALYSES

If children were as infectious as adults, the average number of secondary cases in elementary schools 30 days was 1.9 times higher compared to the base case; if adolescents were half as susceptible as adults, secondary cases in high schools were reduced by a factor of 0.3 (Figure S2, S3). When we repeated our analysis of elementary schools assuming equivalent variation in individual infectiousness for children as for adults, we found more instances of no onward transmission and a lower average number of secondary infections over 30 days, but also slightly larger outbreaks when they occurred (Figures S2, S3). After a single introduction, all types of 2-day schedules (e.g. MW/ThF vs. MT/WTh) led to similar numbers of secondary infections,

similar chance of any secondary infection, and similar numbers of secondary infections among teachers/staff (Figures S2, S2). The benefits of hybrid scheduling generally persisted across increased levels of out-of-school mixing (Figures S5-S6). In order for student and caregiver secondary infections to be greater with an A/B model than a 5-day model, >9 families needed to interact on each day out of school.

DISCUSSION

While in-person education poses some COVID-19 transmission risk, the results of this simulation model underscore that it is possible to offset this with adequate precautions, particularly in elementary schools and when community transmission is well-controlled. In elementary schools with adherence to masking and distancing, our results demonstrate most cases introduced into a school would lead to little or no onward transmission. Nevertheless, if transmission occurs, it may be difficult to link to the school, and even modeled scenarios that commonly lead to no in-school transmission can occasionally generate larger outbreaks in a school community. The risk of this is substantially higher in high schools compared to elementary schools.

While it is difficult to determine the exact risk of these low-probability but high-consequence events, community transmission determines the number of cases that enter a school community and therefore such rare but consequential outbreaks become increasingly likely when local incidence is high. The guidance from the US Centers for Disease Control and Prevention regarding K-12 education highlights population-adjusted incidence as a core indicator (50), and our results provide additional evidence supporting use of this metric for school decisions in the absence of in-school screening or surveillance data. Nevertheless, while schools may decide that in-school transmission risk is too high above certain levels of community incidence, our results do not suggest that K-12 education with mitigation and/or modified schedules, is likely to be a primary driver of sustained community transmission. In the event that community-level $R(t)$ exceeds 1, it may be possible to lower this through targeted closure of high risk venues, and maintain $R(t)$ below 1 while schools remain open.

Our results are compatible with global observations regarding school outbreaks, in which numerous well-studied index cases have produced no or minimal secondary cases, but larger outbreaks have also been recorded, particularly in secondary schools (10,51,52). Teachers and staff tend to be overrepresented in school outbreaks relative to their presence in a school community. At the elementary level, they often represent a third or more of diagnosed secondary cases (7,12). As a substantial fraction of staff may be at high risk of complications, it is important that schools undertake precautions specifically to prevent transmission among staff. These may include holding faculty meetings virtually, mandating masks during all work in shared spaces, and establishing locations for lunches and mask breaks that permit distance and ventilation.

Hospitals have similarly found patterns of staff-to-staff transmission and implemented measures such as reducing opportunities for communal food consumption (53). It is also critical that staff have adequate financial and logistical support to isolate when exhibiting symptoms and to quarantine following COVID-19 exposure, even if exposed outside of the school setting.

We predict that most in-school transmission will occur in the classroom during sustained contact, and consequently interventions which reduce classroom transmission can be highly effective, such as distancing, masking, or reducing class sizes. The latter may be achieved through the addition of more staff and moving into previously unused spaces, by allowing families to opt out of in-person learning, by prioritizing a subset of vulnerable students for limited in-person slots, or through implementation of hybrid scheduling. Some have raised concerns that a hybrid model could paradoxically increase SARS-CoV-2 transmission in schools by leading to greater out-of-cohort mixing on days when children are not physically in school. However, we find that the A/B schedule leads to far fewer infections in the school community than a 5-day schedule, a result that persists even with an assumption of substantial out-of-school mixing between families. Weekly SARS-CoV-2 testing of asymptomatic people (screening) could offer benefits comparable to an A/B schedule, and thereby allow more in-person instruction with less disruption.

Weekly asymptomatic SARS-CoV-2 screening is particularly valuable in light of uncertainty in model parameters and outcomes. There is still debate surrounding relative susceptibility and transmissibility of children compared to adults. In addition, the degree of variation in individual-level infectiousness is uncertain. Such variation can arise from biologic factors associated with viral control, perhaps ACE2 receptor density, or from biologic or behavioral factors that lead to spread of respiratory droplets such as body size, cough strength, and tendency to contain coughs or sneezes (48,54–57). Increases in this “overdispersion” lead to a wider range of possible outcomes when an infection is introduced, including both greater instances of zero in-school transmission as well as larger outbreaks when secondary transmission occurs. This variability in possible outcomes is markedly reduced when hybrid schedules and/or weekly screening are implemented.

Beyond biological uncertainty, it may also be difficult to ascertain local behavioral parameters such as adherence to masking, distancing, and quarantine protocols, which can vary across different settings and over time. Coupled with stochastic uncertainty in outbreak size and a high proportion of subclinical illnesses in school populations, it may be difficult to ensure minimal transmission and detect outbreaks that occur before they spread to the community. In particular, the low probability of clinical disease for infected children means that transmission chains in schools are likely to be only partially observed, and linked cases may be mistakenly classified as isolated introductions. While regular screening can both improve these data and

prevent transmission through early detection, at a minimum, schools should be on alert for signs that an outbreak is brewing undetected, and consider screening of all students and staff in response to the detection of any case without a clearly identified in-school or out-of-school link. Other factors not considered in this work that may affect transmission include changes in parent behavior when children return to school and staff social interactions outside of school.

Last, a new variant of COVID-19 (B.1.1.7) has recently been identified that is approximately 50% more transmissible than the strains that have previously dominated, with possibly larger increases in transmissibility among children than among adults compared to the previous strain (58). This variant has been linked to a large outbreak in a primary school in the Netherlands and triggered a new wave of school closures in Europe (19,59). If this variant spreads throughout the United States, classroom-based efforts at infection prevention such as masking and distancing may be less effective at suppressing attack rates. For instance, schools which have currently achieved attack rates commensurate with “medium uptake of mitigation” may find that the same measures will result in attack rates reflecting “low uptake.” This shift points to the even more urgent need for data and underscores the added value of routine asymptomatic screening.

There are inevitable trade-offs between school disruption, risk of in-school transmission, and resources required to implement in-school prevention approaches. Decisions about the acceptable balance between these trade-offs are best made by school communities. We offer a quantitative perspective on the way that in-school transmissions are likely to occur, as well as the impact that proposed protocols could have on that transmission risk. We emphasize that particularly among young children, schools appear to be ‘mirrors’ of community transmission, rather than ‘amplifiers’ or ‘brakes.’ Thus, a reliable way to ensure a low infection risk in school is to reduce infectious introductions, by keeping community incidence low. However, even when introductions occur, high adherence to in-school prevention measures, preferably complemented by regular asymptomatic testing and teacher vaccination, can also permit return to in-person education with controlled risk of in-school COVID-19 transmission. Local, state, and federal agencies should prioritize these effective interventions that permit the benefits of in-person education while protecting the safety of both students and educators.

FIGURES AND TABLES

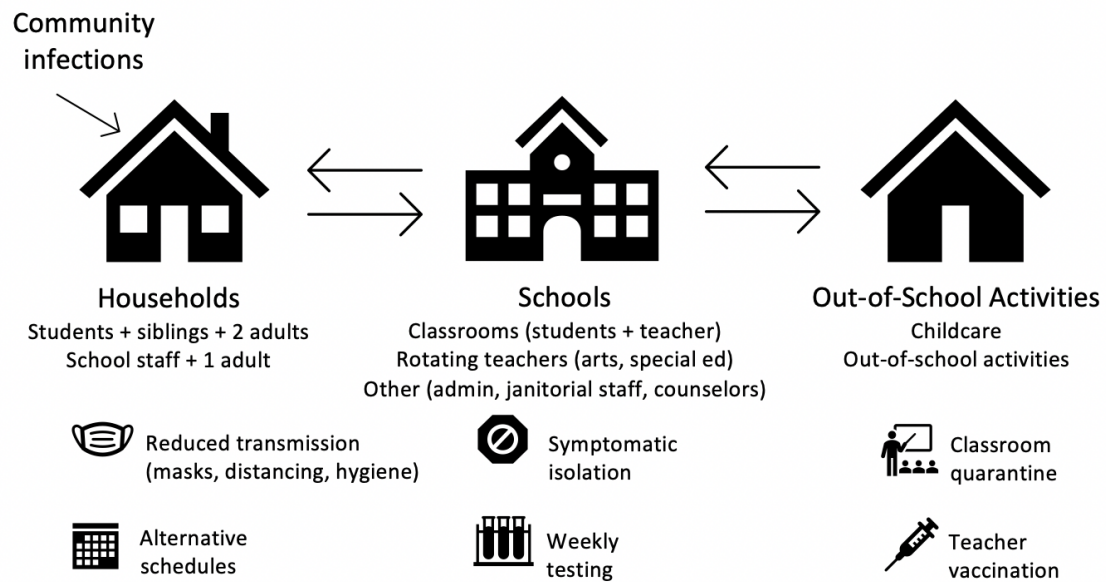


Figure 1: Model diagram. The model includes 3 primary domains: households, schools, and out-of-school social/childcare mixing and incorporates a range of interventions to prevent or reduce transmission.

Table 1: Parameter values.

Parameter	Value	Source
Number of students in elementary school	638	(37)
Average of students per elementary school class	21.3	Consistent with (37), 5 classes per grade
Number of adults who are not primary teachers in elementary schools	30	Estimate via school websites & personal communication
Number of students in high school	1451	(37)
Average of students per high school class	23	Consistent with (37), 16 classes per grade
Number of adults who are not primary teachers in elementary schools	30	Estimate via school websites & personal communication
Probability of asymptomatic disease for adults	0.2	(46)
Probability of asymptomatic disease for children	0.4	(8)
Relative infectiousness of asymptomatic disease	0.5	(46)
Probability of subclinical disease for adults (includes asymptomatic)	0.4	
Probability of subclinical disease for children (includes asymptomatic)	0.8	(60)
Household attack rate	20% (~4% per day)	(38)

Parameter	Value	Source
Classroom attack rate	1/2/3% per day	
Relative attack rate for random school contacts (vs. classroom)	0.13	Approximately 45 minutes
Relative attack rate for random staff contacts (vs. classroom)	2	
Relative attack rate for childcare contacts (vs. classroom)	2	Assuming fewer precautions
Distribution on viral load	Lognormal (.84, .3)/.84	Produces 20%/45% from differences in viral load (47)
Incubation period (from exposure to symptoms, if symptomatic)	Gamma (5.8, 0.95)	Estimated from (61)
Start of infectiousness relative to symptoms	Normal (1.2, 0.4)	Estimated as mean 1.3 days, (63) mean 1.2 days w/lower bound of 95% CI of 3 for start and 2 for peak (64)
Duration of infectiousness	Lognormal (5, 2)	(47)
Duration of isolation and quarantine	10 days	(32)
Teacher vaccination uptake	75%	(1,33)
Vaccination effectiveness	90%	(35), assuming that vaccine blocks infections as well as prevents symptomatic illness (currently unknown)
Test uptake	90%	Assumed
Test sensitivity during infectious period	90%	

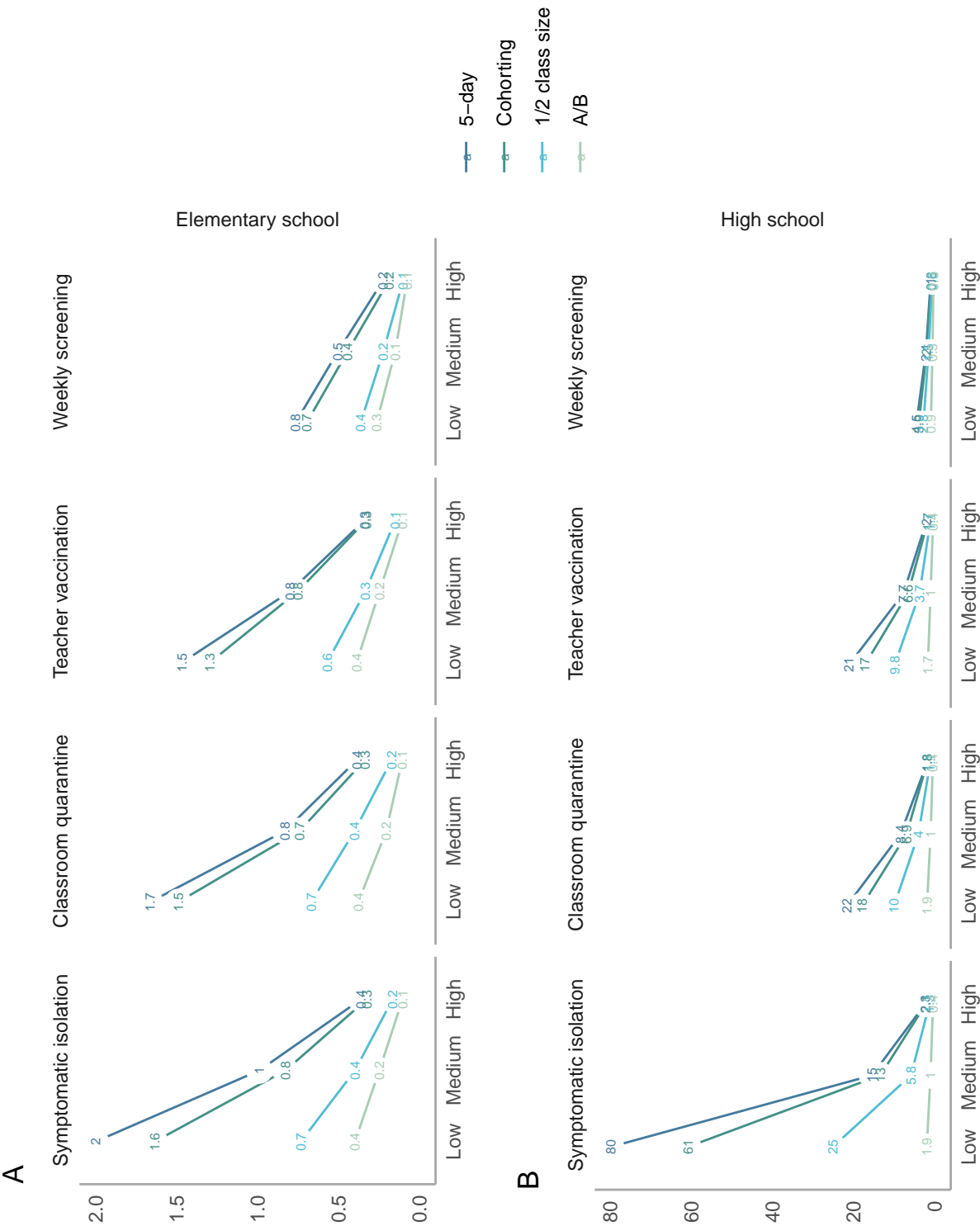


Figure 2: Average number of total secondary transmissions over 30 days (outside of the index case's household) following a single introduction into a school community. (This is not an estimate of R, the effective reproduction number, which is displayed in Figure S8.) These include both transmission directly from the index case, as well as from secondary and tertiary cases. The top panel shows elementary schools, where children are assumed to be less susceptible and less infectious, while the bottom panel shows high schools. Note that axes differ across rows. The x-axes vary the level of prevention measure uptake, with low uptake assuming minimal interventions and high uptake assuming intensive interventions. Line colors correspond to scheduling strategies.

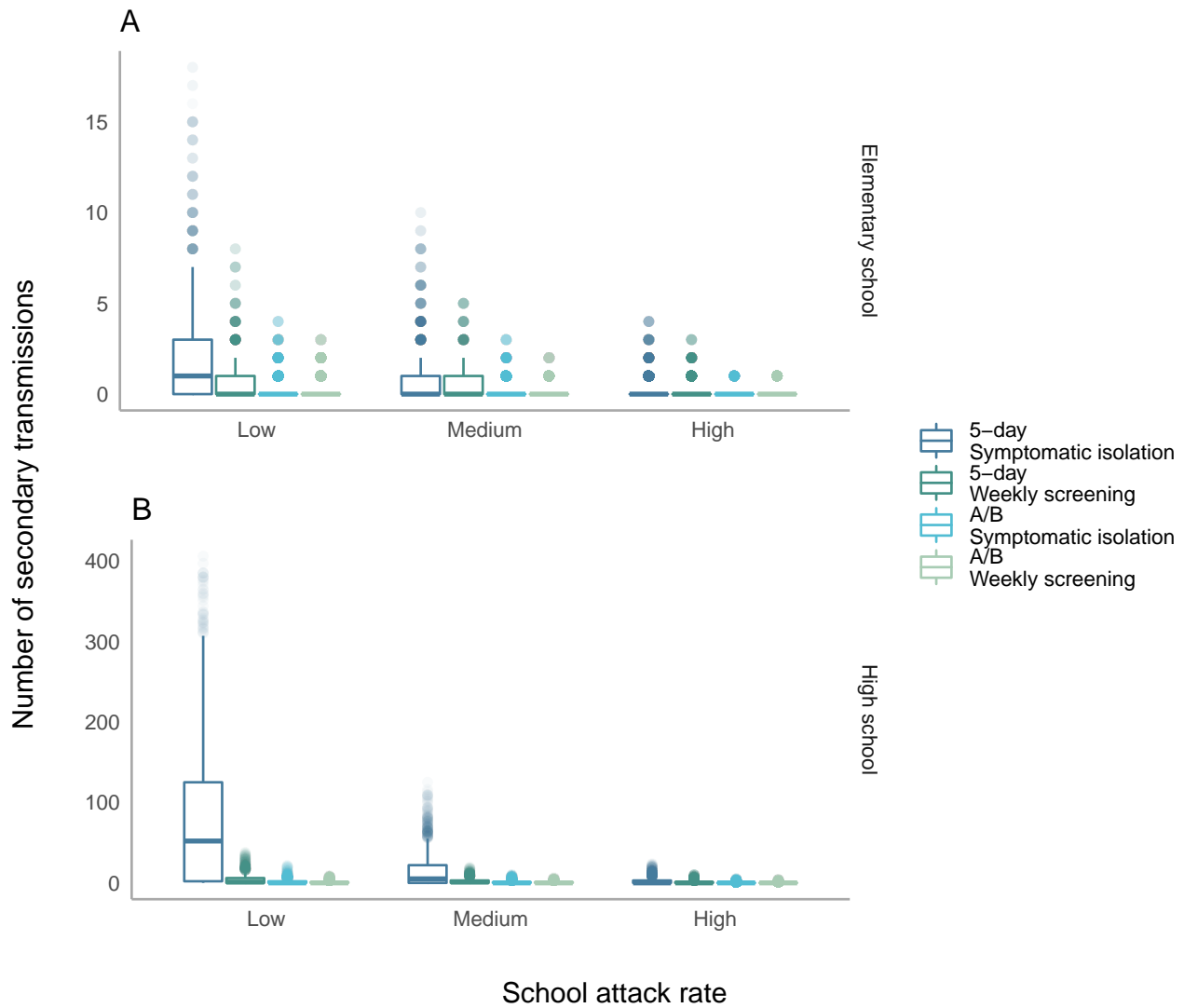


Figure 3: Distribution of secondary transmissions when a single case is introduced. The y-axis displays the number of secondary transmissions (outside of the index case's household) when a case is introduced. Transmissions include both those directly from the index case, as well as those from secondary and tertiary cases. Distributions are truncated at the 99.5th quantile, i.e. all outcomes occur with at least probability $1/200$.

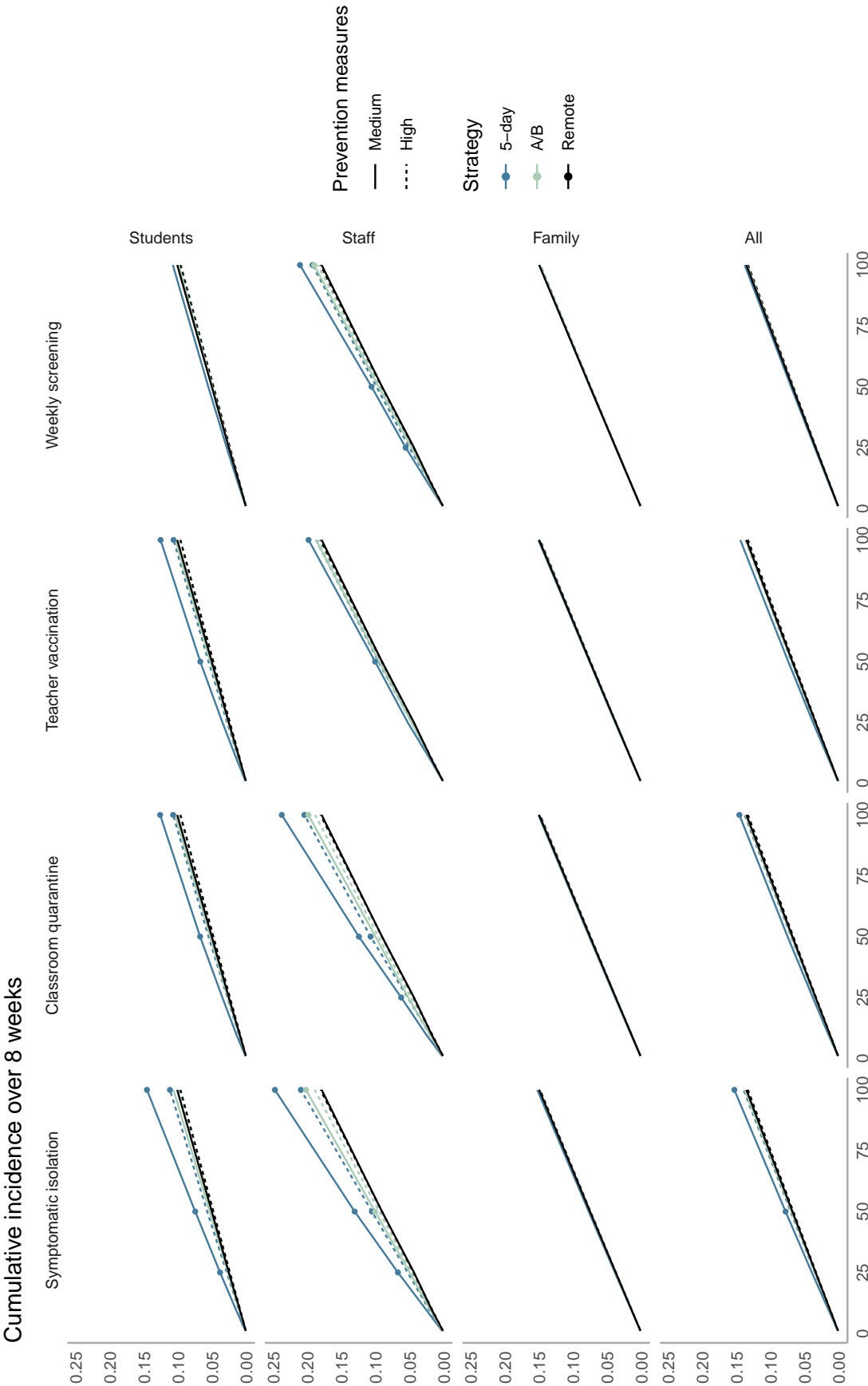


Figure 4: Cumulative incidence over 8 weeks in elementary schools. The x-axis shows the average daily community incidence per 100,000 population. The y-axis shows cumulative incidence over 8 weeks. Columns denote different isolation, quarantine, vaccination, and detection strategies, while rows show different population subgroups. Points are marked for strategies with increased incidence over remote learning that exceeds 1%.

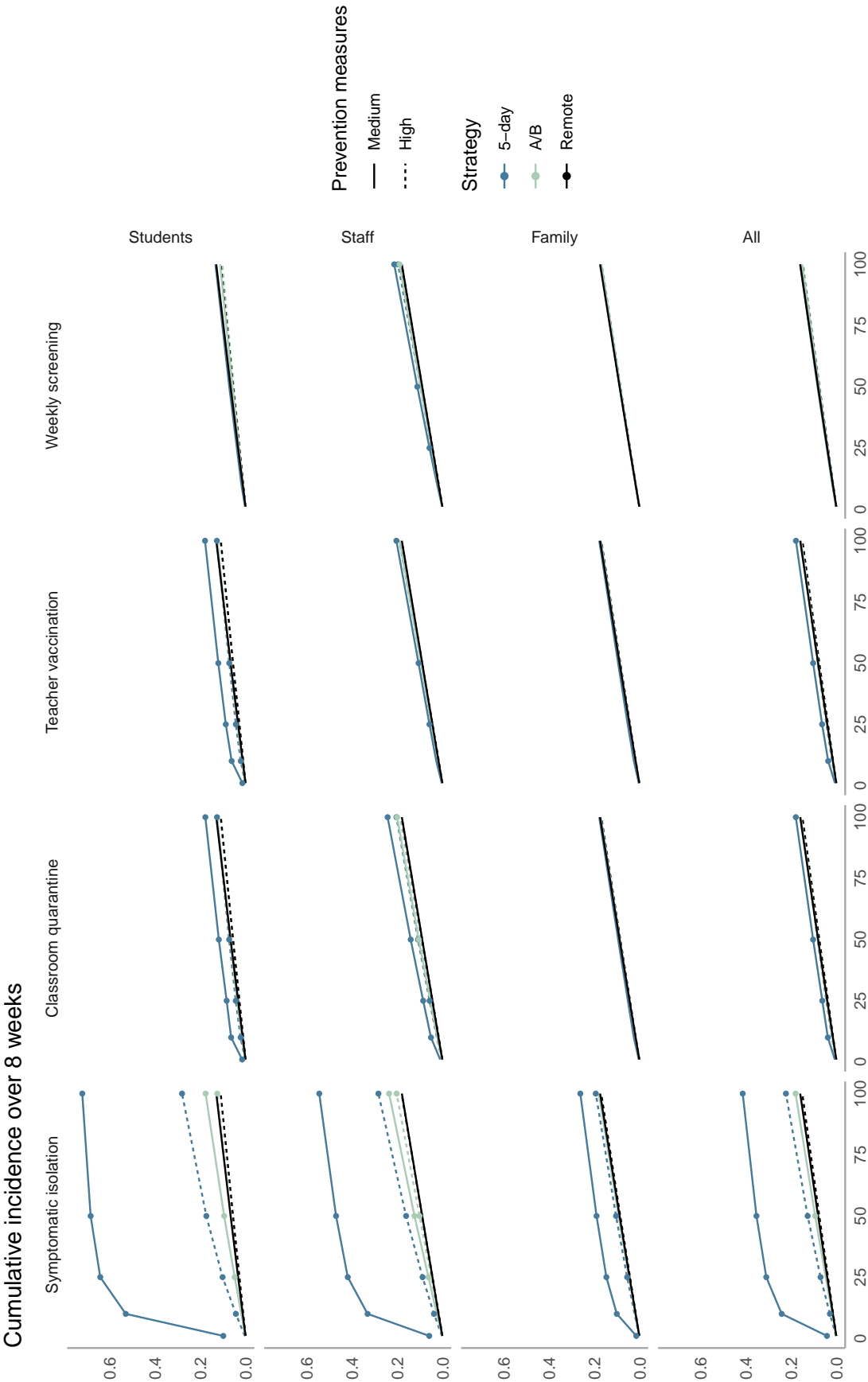


Figure 5: Cumulative incidence over 8 weeks in high schools. The x-axis shows the average daily community incidence per 100,000 population. The y-axis shows cumulative incidence over 8 weeks. Columns denote different isolation, quarantine, vaccination, and detection strategies, while rows show different population subgroups. Points are marked for strategies with increased incidence over remote learning that exceeds 1%.

SUPPLEMENT

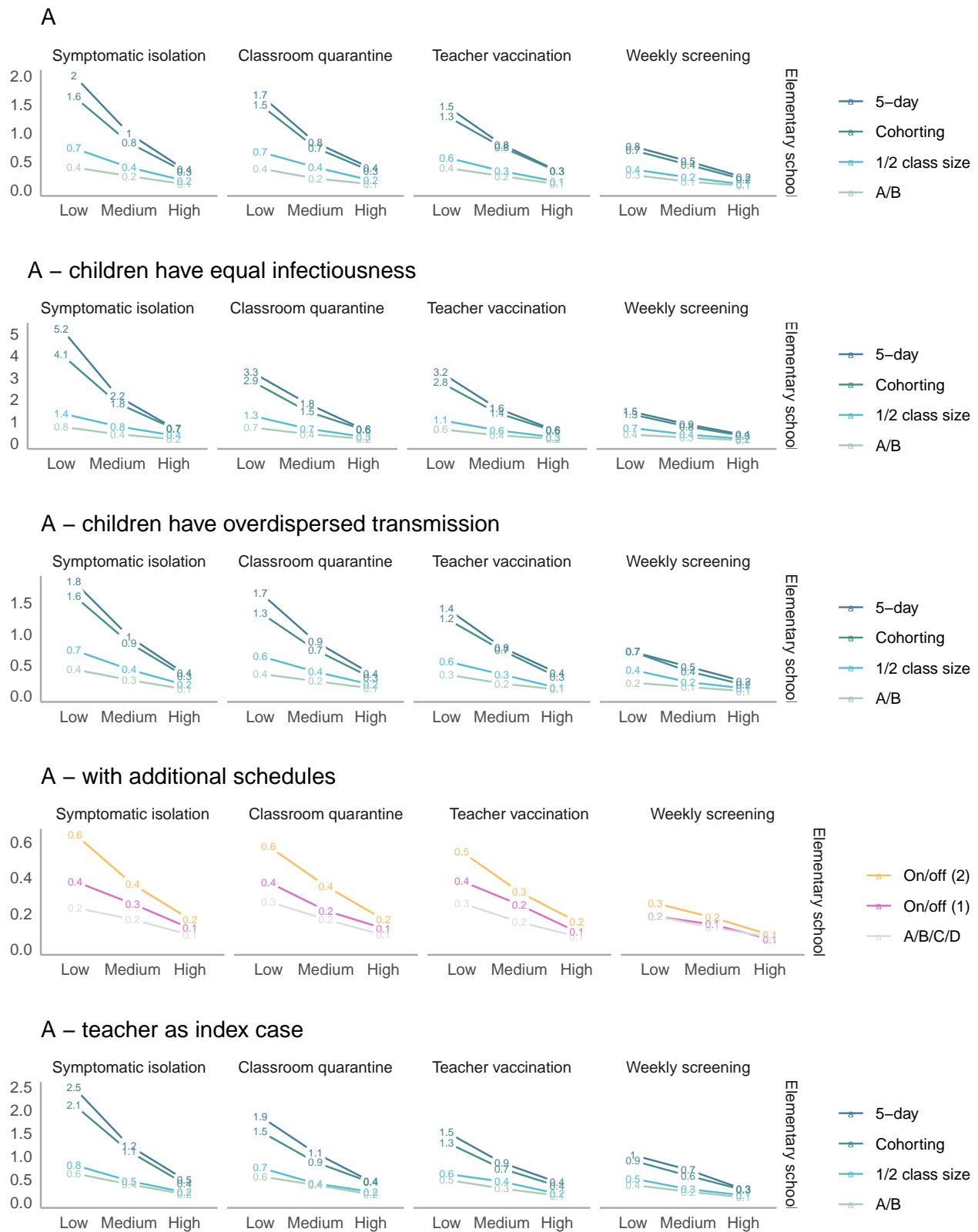


Figure S1: Sensitivity analyses (elementary schools) – average number of total secondary transmissions over 30 days (outside of the index case's household) following a single introduction into a school community. The columns vary the level of prevention measure uptake, with low uptake assuming minimal interventions and high uptake assuming intensive interventions. Line colors correspond to scheduling strategies.

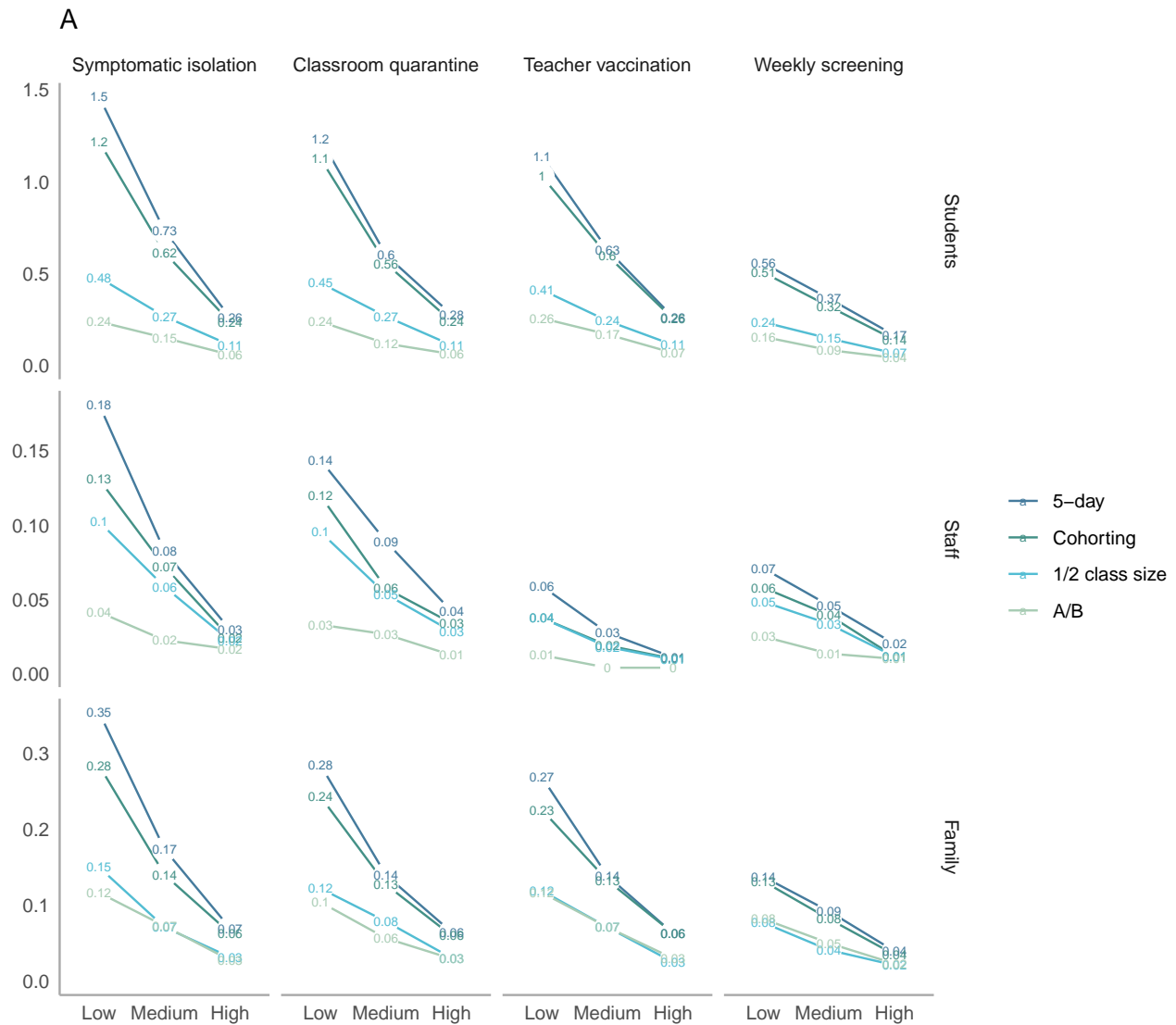


Figure S2: Sensitivity analysis - elementary schools by case type. Average number of total secondary transmissions over 30 days (outside of the index case's household) following a single introduction into an elementary school community. These include both transmission directly from the index case, as well as from secondary and tertiary cases. The x-axis varies the level of prevention measure uptake, with low uptake assuming minimal interventions and high uptake assuming intensive interventions. Line colors correspond to scheduling strategies.

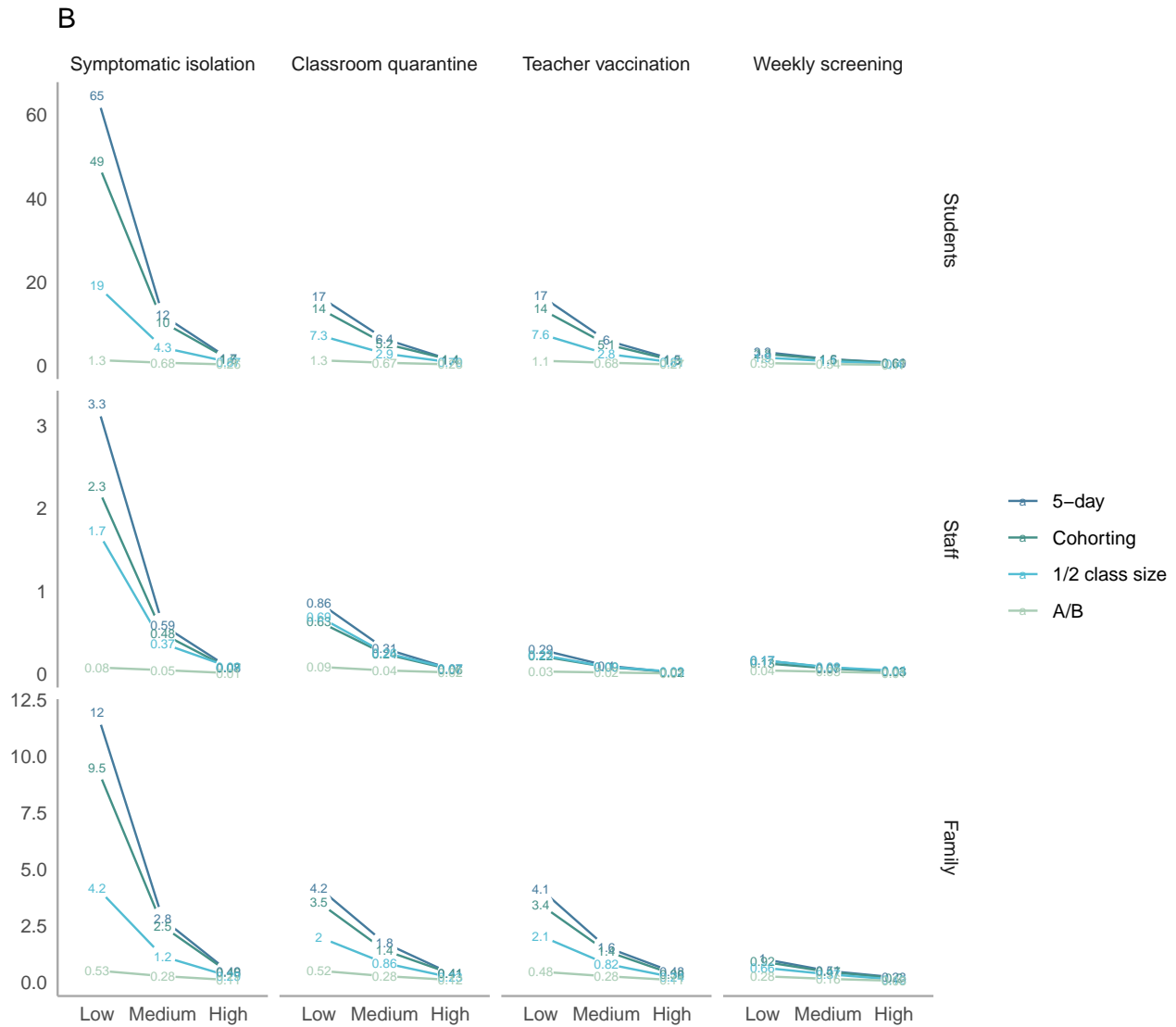


Figure S3: Sensitivity analysis - high schools by case type. Average number of total secondary transmissions over 30 days (outside of the index case's household) following a single introduction into an elementary school community. These include both transmission directly from the index case, as well as from secondary and tertiary cases. The x-axis varies the level of prevention measure uptake, with low uptake assuming minimal interventions and high uptake assuming intensive interventions. Line colors correspond to scheduling strategies.

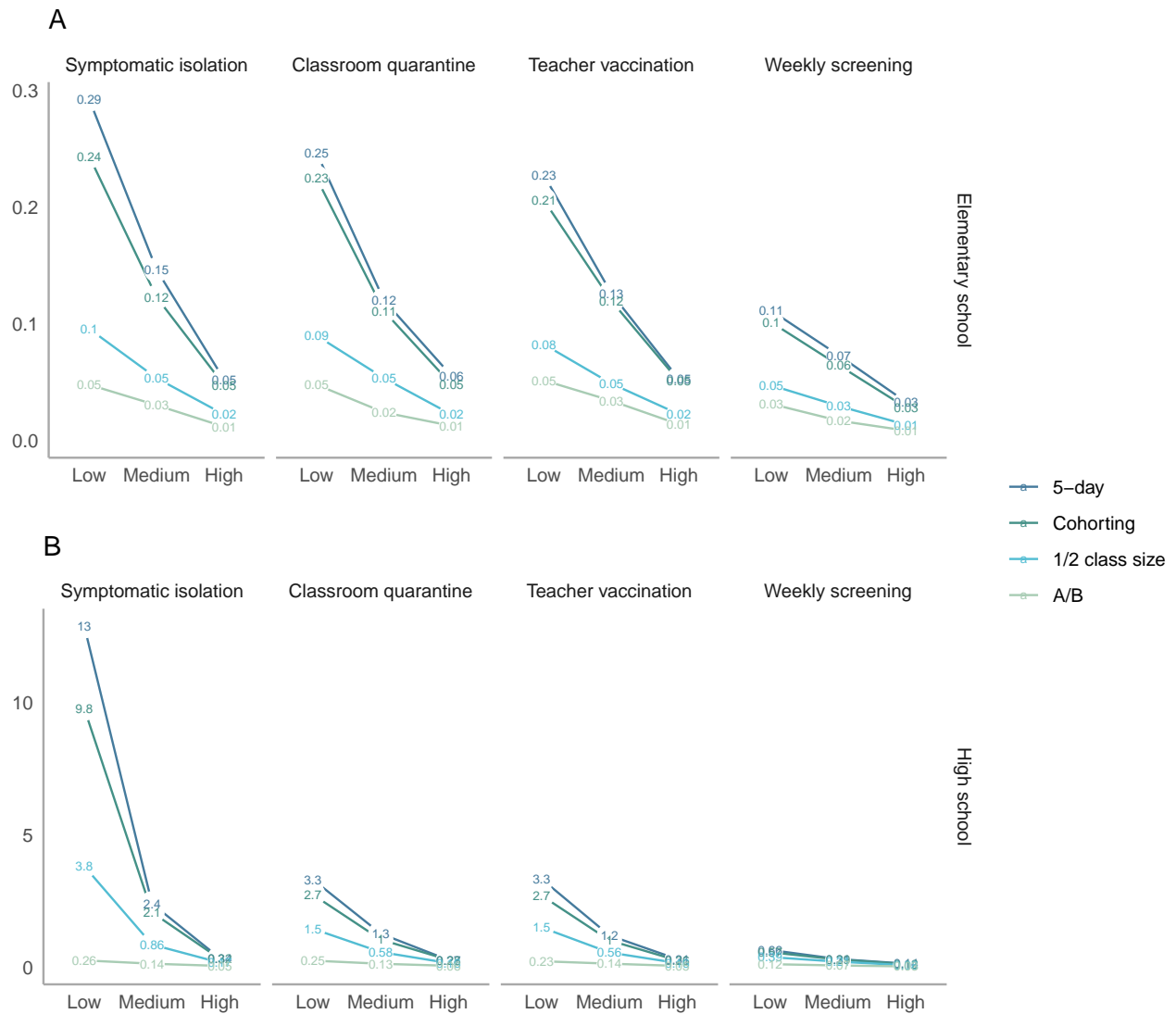
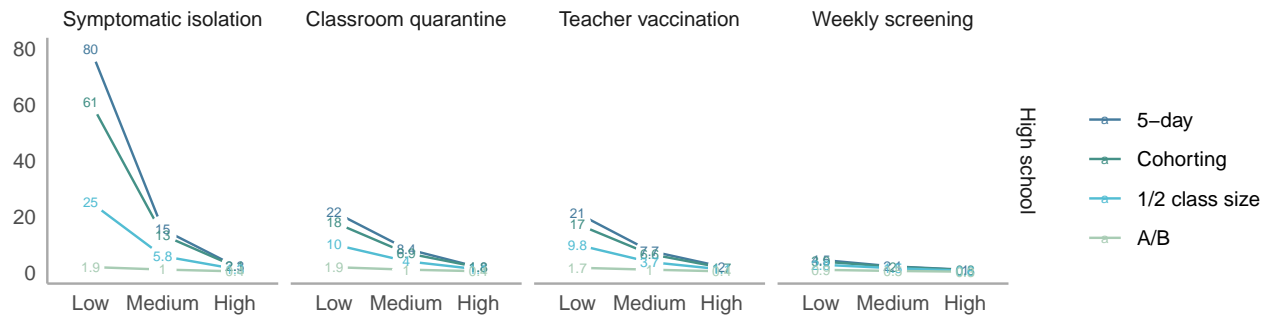
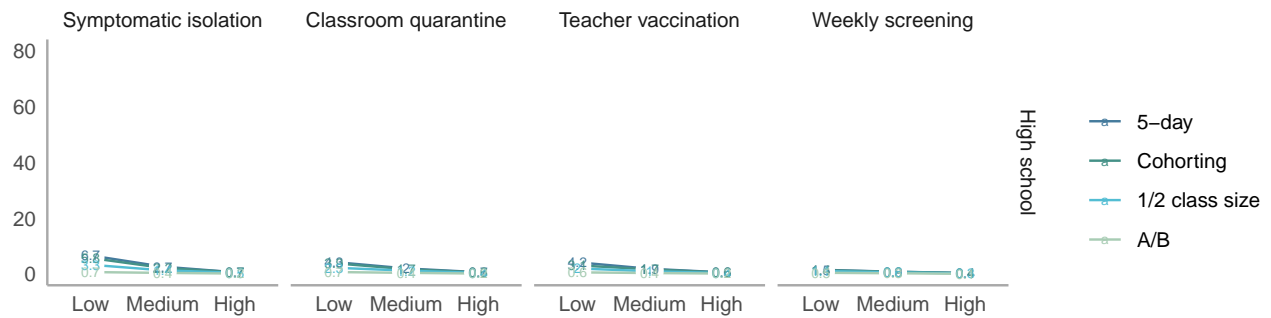


Figure S4: Average number of clinically symptomatic cases in staff and students over 30 days following a single introduction into a school community. These include both transmission directly from the index case, as well as from secondary and tertiary cases. The top panel shows elementary schools, where children are assumed to be less susceptible and less infectious, while the bottom panel shows high schools. Note that axes differ across rows. The x-axes vary the level of prevention measure uptake, with low uptake assuming minimal interventions and high uptake assuming intensive interventions. Line colors correspond to scheduling strategies.

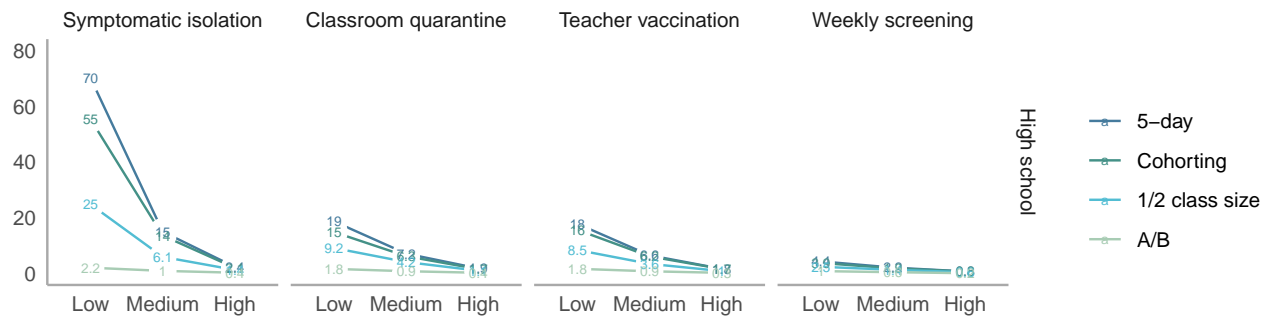
B



B – adolescents less susceptible



B – adolescents have overdispersed transmission



B – with different schedules

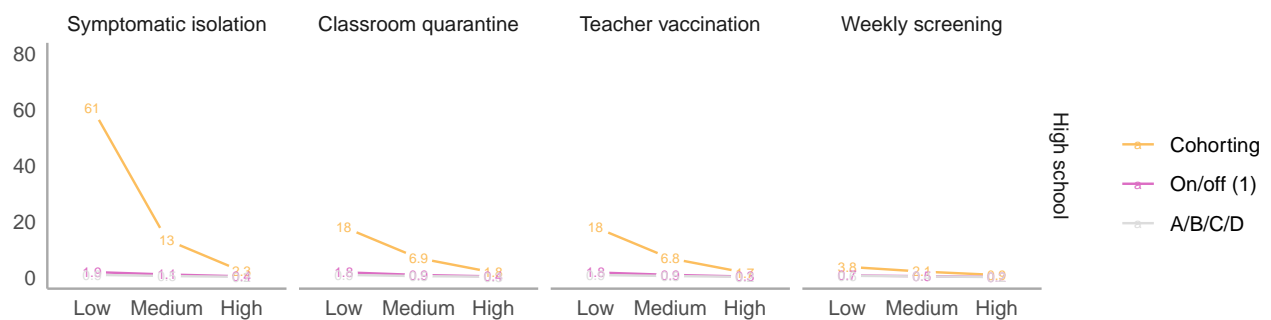


Figure S5: Sensitivity analyses (high schools) – average number of total secondary transmissions over 30 days (outside of the index case's household) following a single introduction into a school community. The columns vary the level of prevention measure uptake, with low uptake assuming minimal interventions and high uptake assuming intensive interventions. Line colors correspond to scheduling strategies.

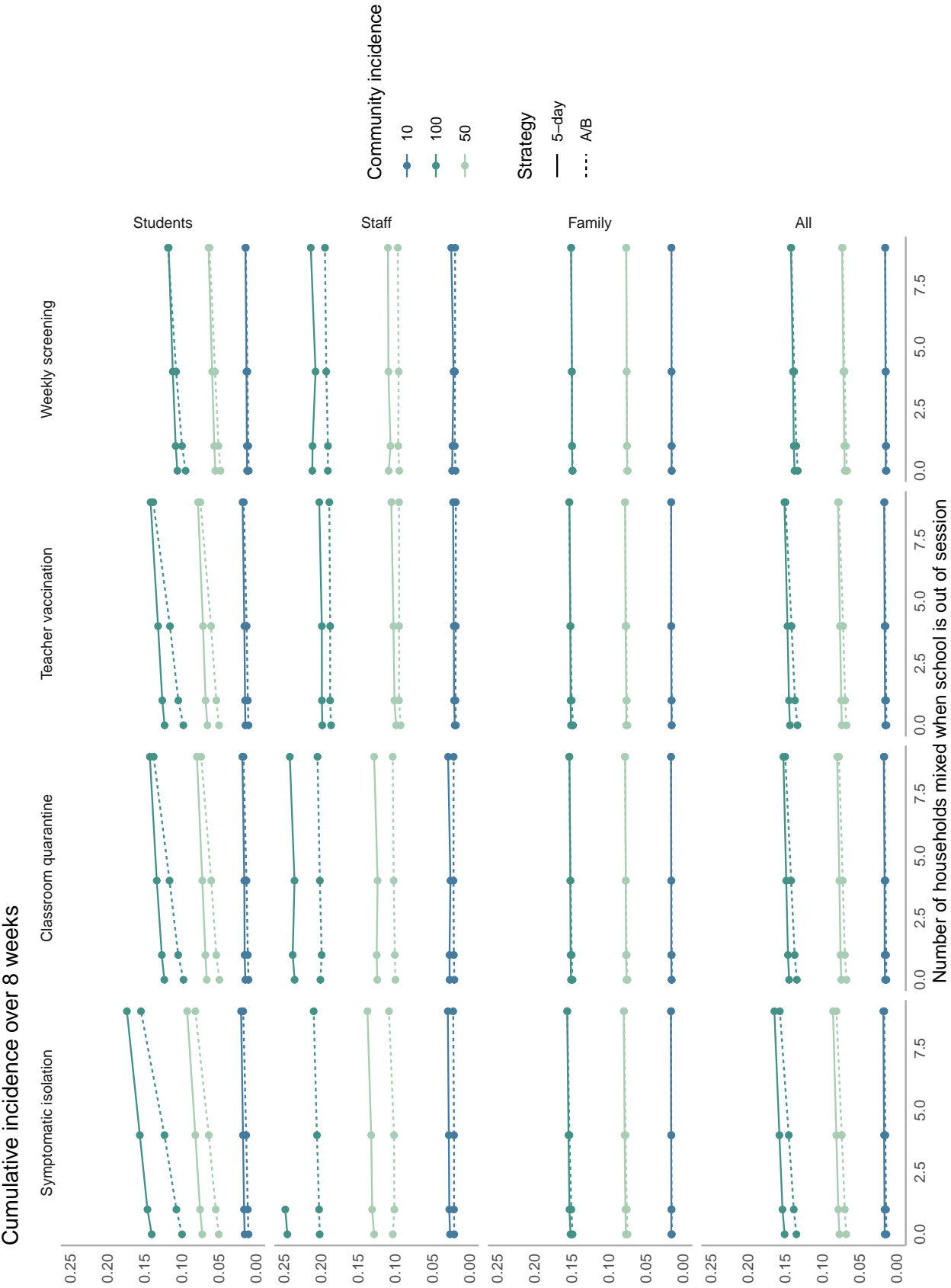


Figure S6: Cumulative incidence over 8 weeks in elementary schools across different levels of out-of-school mixing. The x-axis shows the average daily community incidence per 100,000 population. The y-axis shows cumulative incidence over 8 weeks. Columns denote different isolation, quarantine, and detection strategies, while rows show different population subgroups.

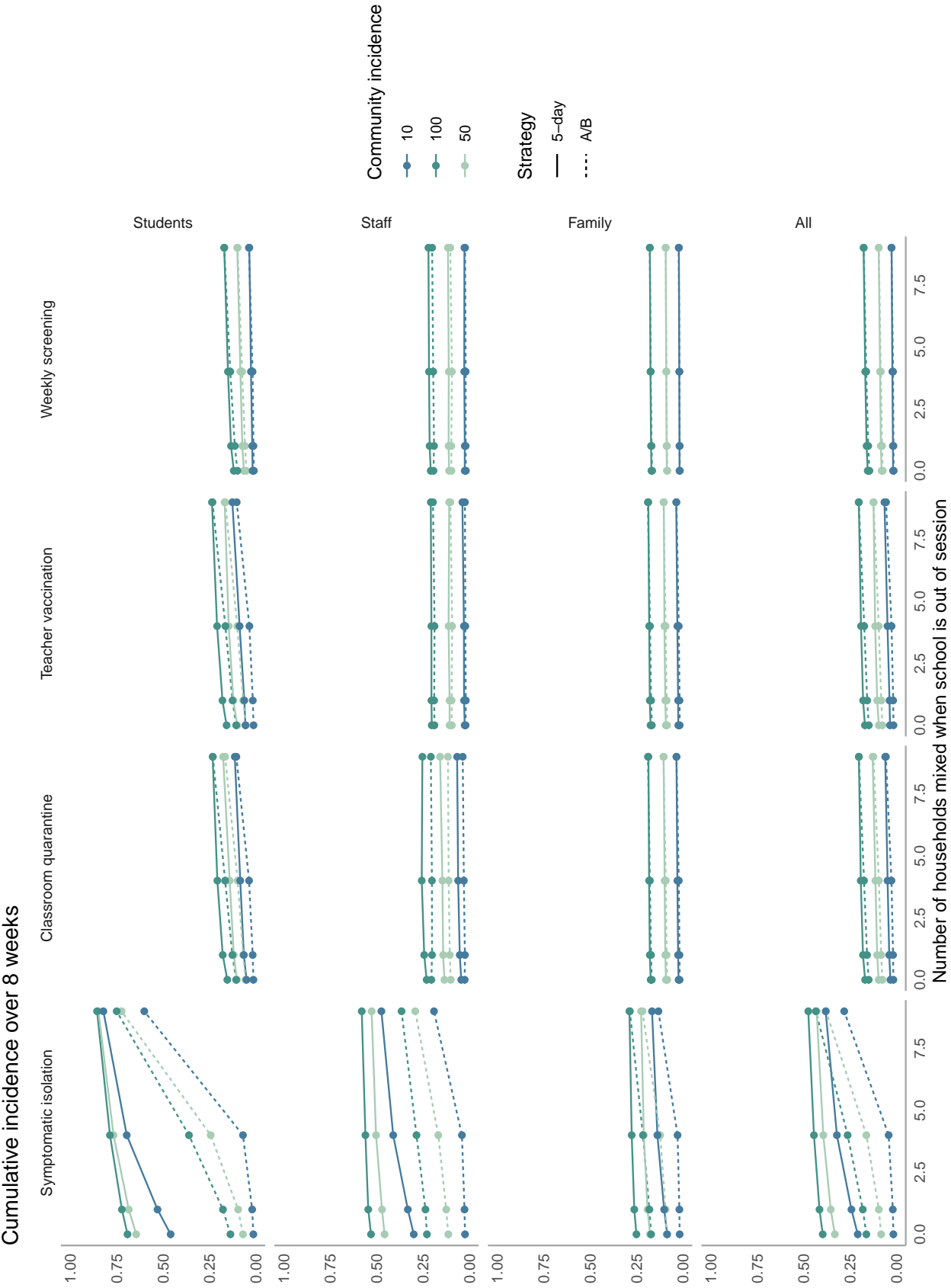


Figure S7: Cumulative incidence over 8 weeks in high schools across different levels of out-of-school mixing. The x-axis shows the average daily community incidence per 100,000 population. The y-axis shows cumulative incidence over 8 weeks. Columns denote different isolation, quarantine, and detection strategies, while rows show different population subgroups.

MODEL

We use a Framework for Reconstructing Epidemiological Dynamics (FRED) to generate household structures (36). For computational simplicity, we used Maryland as a representative state, as sibling structure (the main parameter of interest) did not appear sensitive to location.

Maryland Elementary

Fraction of households with a child of age 1 containing a child of age 2

Age 2	10	0.08	0.12	0.12	0.14	0.06	1
	9	0.09	0.12	0.14	0.05	1	0.05
	8	0.11	0.13	0.05	1	0.05	0.13
	7	0.13	0.06	1	0.05	0.14	0.12
	6	0.05	1	0.06	0.13	0.11	0.11
	5	1	0.05	0.12	0.11	0.08	0.07
		5	6	7	8	9	10
		Age 1					

Connecticut Elementary

Fraction of households with a child of age 1 containing a child of age 2

Age 2	10	0.07	0.11	0.14	0.17	0.04	1
	9	0.1	0.12	0.15	0.05	1	0.04
	8	0.14	0.15	0.06	1	0.04	0.16
	7	0.15	0.07	1	0.07	0.15	0.13
	6	0.07	1	0.06	0.15	0.12	0.09
	5	1	0.07	0.14	0.13	0.09	0.06
		5	6	7	8	9	10
		Age 1					

Mississippi Elementary

Fraction of households with a child of age 1 containing a child of age 2

Age 2	10	0.1	0.1	0.12	0.12	0.07	1
	9	0.12	0.15	0.12	0.08	1	0.07
	8	0.12	0.11	0.07	1	0.07	0.12
	7	0.14	0.08	1	0.07	0.11	0.12
	6	0.08	1	0.09	0.11	0.14	0.1
	5	1	0.08	0.14	0.12	0.11	0.1
		5	6	7	8	9	10
		Age 1					

Texas Elementary

Fraction of households with a child of age 1 containing a child of age 2

Age 2	10	0.09	0.11	0.13	0.15	0.08	1
	9	0.12	0.14	0.14	0.06	1	0.08
	8	0.14	0.15	0.07	1	0.06	0.15
	7	0.15	0.09	1	0.07	0.14	0.13
	6	0.07	1	0.09	0.15	0.14	0.11
	5	1	0.07	0.15	0.14	0.12	0.09
		5	6	7	8	9	10
		Age 1					

Maryland HS

Fraction of households with a child of age 1 containing a child of age 2

Age 2	17	0.13	0.14	0.08	1
	16	0.14	0.06	1	0.08
	15	0.06	1	0.06	0.14
	14	1	0.05	0.12	0.12
		14	15	16	17
		Age 1			

Connecticut HS

Fraction of households with a child of age 1 containing a child of age 2

Age 2	17	0.13	0.14	0.07	1
	16	0.14	0.06	1	0.07
	15	0.06	1	0.06	0.14
	14	1	0.06	0.13	0.13
		14	15	16	17
		Age 1			

Mississippi HS

Fraction of households with a child of age 1 containing a child of age 2

Age 2	17	0.1	0.14	0.09	1
	16	0.14	0.1	1	0.09
	15	0.08	1	0.1	0.13
	14	1	0.08	0.13	0.1
		14	15	16	17
		Age 1			

Texas HS

Fraction of households with a child of age 1 containing a child of age 2

Age 2	17	0.13	0.14	0.08	1
	16	0.16	0.08	1	0.08
	15	0.08	1	0.08	0.14
	14	1	0.08	0.17	0.13
		14	15	16	17
		Age 1			

COMPARISON TO OBSERVED OUTBREAKS

A number of factors make formal calibration challenging for this paper. First, most data collection has been ad hoc, with some sources biased toward reporting large outbreaks and others toward high-mitigation schools who voluntarily collect and report data. Without comprehensive data on school mitigation efforts, interpretation can be challenging, as our results emphasize that a large range of outcomes is possible due to both variation in parameters and stochastic uncertainty. Second, testing practices vary, affecting reporting procedures. Last, the definition of “contact” used to construct attack rates also varies, with some sources including even brief contacts while others require more sustained interaction (e.g. >15 minutes without a mask). Overall, this section highlights why we believe our results to be generally consistent with observed data as well as possible limitations for future study. We emphasize that substantial uncertainty persists, and that weekly screening or other surveillance is one of the best tools available for understanding a specific context as well as early detection of outbreaks.

DIFFERENCES IN INFECTIOUSNESS AND SUSCEPTIBILITY BY AGE

In our model, we assume that young children (10 and under) are both less susceptible and less infectious than adults. To inform these assumptions, we used a meta-analysis on child susceptibility on those under 18-20 (43), which was consistent with best-fit model estimates (67) and another on child infectiousness (68). We also used a number of contact tracing studies suggesting not just a difference between children and adults but also an age gradient in susceptibility and infectiousness. These included a study from B’nei Brak, Israel on household infections (Figure 4) (44) and two studies from France contrasting minimal elementary school outbreaks (despite introductions) with a larger high school cluster in an area with early COVID-19 exposure (8,45). Limited data from Iceland, with comprehensive contact tracing and sequencing, suggested a similar difference between young children and adolescents in infectiousness (69). Contact tracing data from South Korea on infectiousness, while more difficult to interpret due to concurrent exposures among household members and high PPE usage of guardians of infected children (70,71), was not inconsistent with this finding (31).

We focused on data from contact tracing studies that used comprehensive testing of contacts because these were less likely to be biased. In particular, we did not want to interpret evidence that children were rarely identified as index cases (68) or had lower seroprevalence in some contexts as evidence that they were less susceptible or infectious (72). These differences could have been driven by the fact that children are less likely to have symptoms and be tested and/or that their contacts were markedly reduced by school closures early in the pandemic. While household contact tracing studies with comprehensive testing avoided this

issue, they could have other biases. Children are unlikely to be caretakers of sick individuals, and may have been shielded in houses with transmission, particularly in the midst of an unprecedented pandemic. Nevertheless, in general, higher household attack rates in children have been observed for seasonal influenza and H1N1, making lower estimates for COVID-19 particularly notable (73–78).

These findings nevertheless cannot differentiate between biological explanations for lower susceptibility in younger children (e.g., lower density of ACE2 receptor) and behavioral ones (e.g., easier to restrict socialization). In addition, some studies suggest higher susceptibility and/or infectiousness of young children than we include in our base case (79–83). While we model these possibilities in sensitivity analyses, the bulk of evidence on well-studied school outbreaks has pointed to important distinctions between elementary and high school-aged children. For example, when Israel experienced significant outbreaks upon return to school in the early summer, there was a significant outbreak of 178 cases in a middle/high school (concentrated in grades 7–9), but elementary school outbreaks were generally reported as smaller (e.g. 33 cases) (10,18). This difference was also apparent in informal databases, with high schools largely responsible for outbreaks of more than 50 people (e.g. in New Zealand and the US prior to social distancing and Australia, where a school outbreak was reported to be driven by high schoolers) (84–86). In the Netherlands, a health official was quoted saying that significant outbreaks occurred mainly in high schools and universities prior to an elementary school outbreak with B.1.1.7 (59). Exceptions often included significant outbreaks among teachers (e.g., in Chile (12) and Singapore (87)).

SECONDARY CASES

Our results are broadly consistent a few key features of observed data. First, well-studied cases have led to no or minimal outbreaks in a number of settings. In passive surveillance from the United Kingdom during the summer, in-school transmission was identified from 39% of index cases in secondary schools and 26% of cases in primary schools “in the context of small class or bubble sizes, half empty schools, and extensive hygiene measures.” This similar to what we predicted with an A/B model in secondary schools with medium mitigation (36%) and a low mitigation scenario in elementary schools (23%). No onward transmission was found in Singapore or Ireland, each with 3 seed cases (9,88). In Rhode Island childcare settings, which had small class sizes, onward transmission was documented in 4/29 index cases (14%), consistent with 1/2 class size scenario and high mitigation (12%) (51).

Second, several data sources showed signs of overdispersion with the possibility of large outbreaks alongside cases without apparent transmission. In the Rhode Island example, one outbreak involved 10 cases among contacts (10 children, four staff members, and one parent); in another study from Australia, 9 cases in early

childhood education centers led to no onward transmission while one led to 13 infections (7). If these were indeed caused by a single index case, the level of overdispersion in our model may not capture such extreme outcomes.

Last, teachers were often overrepresented in outbreaks even in well-studied outbreaks, with 16% of staff and 10% of students having antibodies in a Chilean outbreak across multiple school levels (12). In the Australian study, adults comprised 8/18 of secondary cases identified (7).

FREQUENCY OF INFECTIONS AND SUBCLINICAL INFECTIONS IN CHILDREN COMPARED TO ADULTS

Two consequential predictions from our model are that both the rate of infections in children is lower and the rate of underdiagnosis is lower in children. With these combined, we expect to see fewer cases in children, but a smaller relative difference when comprehensive surveillance and/or random testing is conducted. While some well-studied cases in schools have led to minimal onward transmission (9), in several studies, children have had higher rates of COVID-19 compared to adults such contexts. For example, in passive surveillance from the United Kingdom, staff had more than 4 times the COVID incidence of students per 100,000; however in random testing, observed prevalence was equal among students and staff ?? In elementary schools in New York in fall 2020, schools reported fewer cases in elementary school students than in staff, but in random surveillance testing from the same period in New York City (manually extracted and analyzed by others), prevalence was equal in students and staff (89,90). While these two modes are difficult to compare directly as people who self-isolate for symptoms are not present for random testing, the contrast remains striking. Less systematically, a major outbreak in an Israeli high school was detected from wide-scale testing after observing 2 unlinked community cases (10), and the first Ontario school to participate in voluntary mass asymptomatic screening closed after uncovering a substantial number of previously undetected cases (91).

EFFECTIVE REPRODUCTION NUMBER

In Figure S8, we display the effective reproduction number associated with different scenarios. One modeling study estimated that from August through October 2020, there was an average effective reproduction number of 0.54 [0.44-0.62] for children 0-9 and 0.75 [0.59-0.89] for children 10-19 (92). As noted in the introduction of this article, school openings varied considerably across the country, but both of these fall within our range of estimates. For elementary schools, it is consistent with full opening and high mitigation, hybrid opening or limited attendance and medium mitigation, or some combination of these with remote school. For high schools, it is most consistent with high mitigation limited attendance or a hybrid model or some combination

of these with remote school.

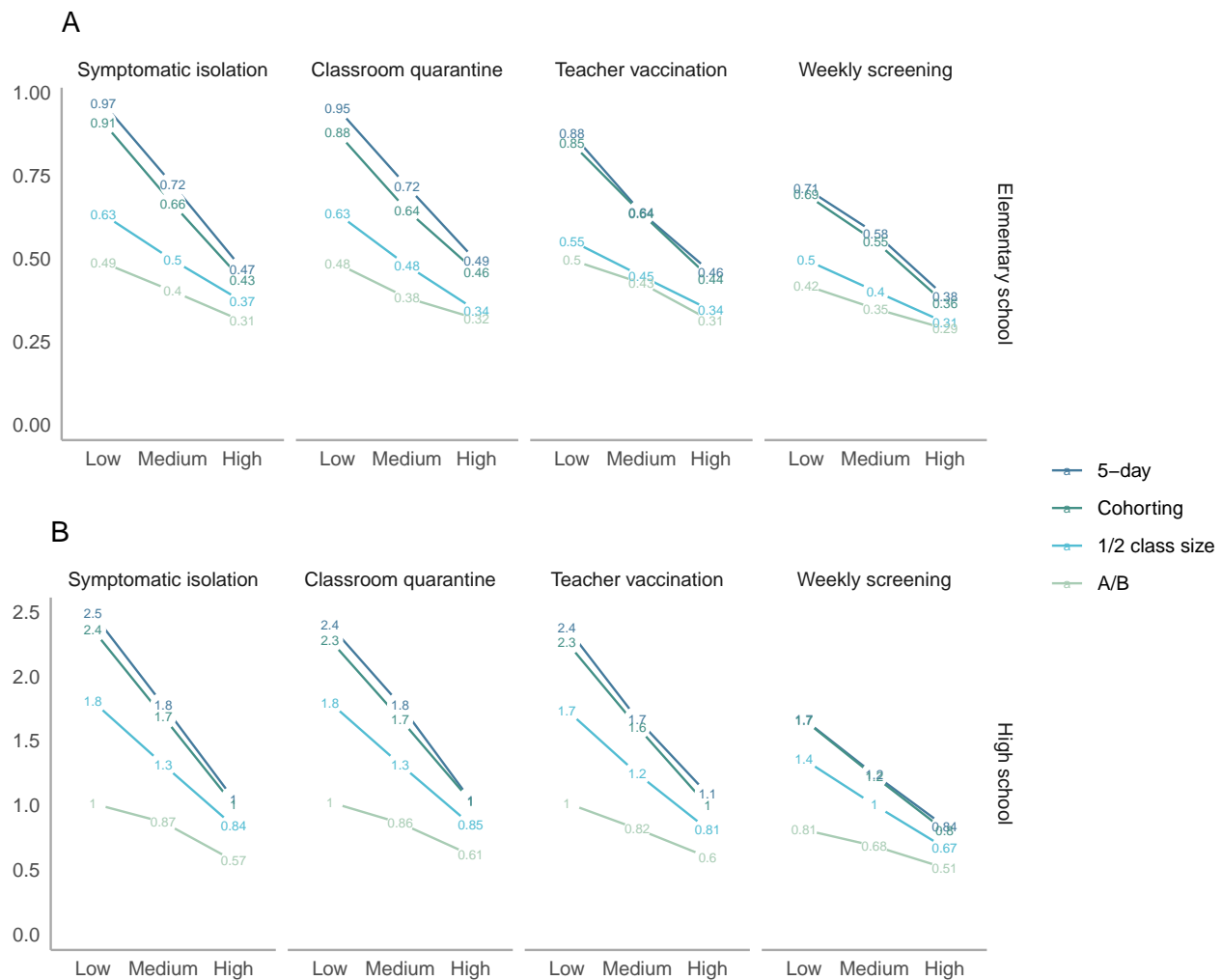


Figure S8: Effective reproduction number, defined as average secondary transmissions following a single introduction into the school community. The top panel shows elementary schools, where children are assumed to be less susceptible and less infectious, while the bottom panel shows high schools. Note that axes differ across rows. The x-axes vary the level of prevention measure uptake, with low uptake assuming minimal interventions and high uptake assuming intensive interventions. Line colors correspond to scheduling strategies.

POPULATION-LEVEL STUDIES

Several population level studies have explored the relationship between school reopening and community incidence. While we do not directly measure community incidence, two recent studies using quasi-experimental data to study the impact of school reopening on transmission were consistent with our observation that there is a higher (or at least inconclusive) risk of increased community transmission following school reopenings when initial transmission was high [(93); goldhaber_what_2020], contrasted to tight null effects when initial transmission was lower.

REFERENCES

1. Map: Coronavirus and School Closures - Education Week. Education Week [Internet]. 2020 Mar [cited 2020 Jun 11]; Available from: <https://www.edweek.org/ew/section/multimedia/map-coronavirus-and-school-closures.html>
2. K-12 School Opening Tracker [Internet]. [cited 2020 Nov 25]. Available from: <https://cai.burbio.com/school-opening-tracker/>
3. More poor, Black and Hispanic students in Central Florida opt for online learning Special Report - Orlando Sentinel [Internet]. [cited 2020 Nov 25]. Available from: <https://www.orlandosentinel.com/news/education/os-ne-race-poverty-school-online-in-person-20201105-wco4evjpxzcbjnx5sgkdsatymi-story.html>
4. Shapiro E. When New York City Schools Reopen, About 700,000 Students Won't Be There. The New York Times [Internet]. 2020 Nov [cited 2020 Nov 25]; Available from: <https://www.nytimes.com/2020/11/20/nyregion/nyc-schools-reopening-coronavirus.html>
5. Shapiro E. How de Blasio Backed Himself Into a Corner on Closing Schools. The New York Times [Internet]. 2020 Nov [cited 2020 Nov 25]; Available from: <https://www.nytimes.com/2020/11/24/nyregion/deblasio-school-reopening.html>
6. Combating COVID-19 [Internet]. President-Elect Joe Biden. [cited 2020 Nov 25]. Available from: <https://buildbackbetter.gov/priorities/covid-19/>
7. Macartney K, Quinn HE, Pillsbury AJ, Koirala A, Deng L, Winkler N, et al. Transmission of SARS-CoV-2 in Australian educational settings: A prospective cohort study. The Lancet Child & Adolescent Health [Internet]. 2020 Aug [cited 2020 Sep 17];0(0). Available from: [https://www.thelancet.com/journals/lanchi/article/PIIS2352-4642\(20\)30251-0/abstract](https://www.thelancet.com/journals/lanchi/article/PIIS2352-4642(20)30251-0/abstract)
8. Fontanet A, Grant R, Tondeur L, Madec Y, Grzelak L, Cailleau I, et al. SARS-CoV-2 infection in primary schools in northern France: A retrospective cohort study in an area of high transmission. medRxiv [Internet]. 2020 Jun [cited 2020 Aug 1];2020.06.25.20140178. Available from: <https://www.medrxiv.org/content/10.1101/2020.06.25.20140178v2>
9. Yung CF, Kam K-q, Nadua KD, Chong CY, Tan NWH, Li J, et al. Novel Coronavirus 2019 Transmission Risk in Educational Settings. Clinical Infectious Diseases [Internet]. [cited 2020 Nov 25]; Available from: <https://academic.oup.com/cid/advance-article/doi/10.1093/cid/ciaa794/5862649>

10. Stein-Zamir C, Abramson N, Shoob H, Libal E, Bitan M, Cardash T, et al. A large COVID-19 outbreak in a high school 10 days after schools' reopening, Israel, May 2020. *Eurosurveillance* [Internet]. 2020 Jul [cited 2020 Sep 17];25(29):2001352. Available from: <https://www.eurosurveillance.org/content/10.2807/1560-7917.ES.2020.25.29.2001352>
11. Iowa COVID-19 Tracker – Please wear a mask. Please social distance. [Internet]. [cited 2020 Nov 25]. Available from: <https://iowacovid19tracker.org/>
12. Torres JP, Piñera C, De La Maza V, Lagomarcino AJ, Simian D, Torres B, et al. SARS-CoV-2 antibody prevalence in blood in a large school community subject to a Covid-19 outbreak: A cross-sectional study. *Clinical Infectious Diseases* [Internet]. [cited 2020 Sep 17]; Available from: <https://academic.oup.com/cid/advance-article/doi/10.1093/cid/cia955/5869860>
13. Auger KA, Shah SS, Richardson T, Hartley D, Hall M, Warniment A, et al. Association Between Statewide School Closure and COVID-19 Incidence and Mortality in the US. *JAMA* [Internet]. 2020 Sep [cited 2020 Nov 25];324(9):859. Available from: <https://jamanetwork.com/journals/jama/fullarticle/2769034>
14. Haug N, Geyrhofer L, Londei A, Dervic E, Desvars-Larrive A, Loreto V, et al. Ranking the effectiveness of worldwide COVID-19 government interventions. *Nature Human Behaviour* [Internet]. 2020 Nov [cited 2020 Nov 25];1–10. Available from: <https://www.nature.com/articles/s41562-020-01009-0>
15. Dong Y, Mo X, Hu Y, Qi X, Jiang F, Jiang Z, et al. Epidemiology of COVID-19 Among Children in China. *Pediatrics* [Internet]. 2020 Jun [cited 2020 Jun 11];145(6). Available from: <https://pediatrics.aappublications.org/content/145/6/e20200702>
16. Reopening schools in Denmark did not worsen outbreak, data shows. *Reuters* [Internet]. 2020 May [cited 2020 Jun 11]; Available from: <https://www.reuters.com/article/us-health-coronavirus-denmark-reopening-idUSKBN2341N7>
17. Pancevski B, Dandanell N. Is It Safe to Reopen Schools? These Countries Say Yes. *Wall Street Journal* [Internet]. 2020 May [cited 2020 Jun 11]; Available from: <https://www.wsj.com/articles/is-it-safe-to-reopen-schools-these-countries-say-yes-11590928949>
18. Vogel G, Couzin-Frankel Nov. 18 J, 2020, Pm 5. As COVID-19 soars in many communities, schools attempt to find ways through the crisis [Internet]. *Science AAAS*. 2020 [cited 2020 Nov 25]. Available from: <https://www.sciencemag.org/news/2020/11/covid-19-soars-many-communities-schools-attempt-find-ways-through-crisis>

19. Staton B, Chazan G, Hall B, Chaffin J. European capitals follow UK with school closures as virus surges [Internet]. 2021 [cited 2021 Jan 22]. Available from: <https://www.ft.com/content/8121ca0a-4d96-4cf5-b5df-a73adc16a606>
20. Polikoff DS Anna Saavedra. Surveys show things are better for students than they were in the spring—or do they? [Internet]. Brookings. 2020 [cited 2020 Nov 20]. Available from: <https://www.brookings.edu/blog/brown-center-chalkboard/2020/11/18/surveys-show-things-are-better-for-students-than-they-were-in-the-spring-or-do-they/>
21. Gaudiano N. Missing: Millions of students [Internet]. POLITICO. [cited 2020 Nov 25]. Available from: <https://politi.co/3dVrNxxg>
22. Leeb RT. Mental Health–Related Emergency Department Visits Among Children Aged 18 Years During the COVID-19 Pandemic — United States, January 1–October 17, 2020. MMWR Morbidity and Mortality Weekly Report [Internet]. 2020 [cited 2020 Nov 30];69. Available from: <https://www.cdc.gov/mmwr/volumes/69/wr/mm6945a3.htm>
23. Hess A. Widespread school closures mean 30 million kids might go without meals [Internet]. CNBC. 2020 [cited 2020 Aug 1]. Available from: <https://www.cnbc.com/2020/03/14/widespread-school-closures-mean-30-million-kids-might-go-without-meals.html>
24. Baron EJ, Goldstein EG, Wallace CT. Suffering in Silence: How COVID-19 School Closures Inhibit the Reporting of Child Maltreatment [Internet]. Rochester, NY: Social Science Research Network; 2020 Jul [cited 2020 Aug 1]. Report No.: ID 3601399. Available from: <https://papers.ssrn.com/abstract=3664803>
25. Bueno C. Bricks and Mortar vs. Computers and Modems: The Impacts of Enrollment in K-12 Virtual Schools [Internet]. Rochester, NY: Social Science Research Network; 2020 Jul [cited 2020 Aug 1]. Report No.: ID 3642969. Available from: <https://papers.ssrn.com/abstract=3642969>
26. Lempel H, Hammond R, Epstein J. Costs of School Closure [Internet]. Brookings Institute; 2009 Sep [cited 2020 Aug 1]. Available from: https://www.brookings.edu/wp-content/uploads/2016/06/0930_school_closure_presentation.pdf
27. Soland J, Kuhfeld M, Tarasawa B, Johnson A, Ruzek E, Liu J. The impact of COVID-19 on student achievement and what it may mean for educators [Internet]. Brookings. 2020 [cited 2020 Aug 1]. Available from: <https://www.brookings.edu/blog/brown-center-chalkboard/2020/05/27/the-impact-of-covid-19-on-student-achievement-and-what-it-may-mean-for-educators/>

28. Cohen JA, Mistry D, Kerr CC, Klein DJ. Schools are not islands: Balancing COVID-19 risk and educational benefits using structural and temporal countermeasures [Internet]. *Epidemiology*; 2020 Sep [cited 2020 Sep 17]. Available from: <http://medrxiv.org/lookup/doi/10.1101/2020.09.08.20190942>
29. Espana G, Cavany S, Oidtman RJ, Barbera C, Costello A, Lerch A, et al. Impacts of K-12 school reopening on the COVID-19 epidemic in Indiana, USA. *medRxiv* [Internet]. 2020 Sep [cited 2020 Sep 17];2020.08.22.20179960. Available from: <https://www.medrxiv.org/content/10.1101/2020.08.22.20179960v2>
30. Head JR, Andrejko K, Cheng Q, Collender PA, Phillips S, Boser A, et al. The effect of school closures and reopening strategies on COVID-19 infection dynamics in the San Francisco Bay Area: A cross-sectional survey and modeling analysis. *medRxiv* [Internet]. 2020 Aug [cited 2020 Sep 17];2020.08.06.20169797. Available from: <https://www.medrxiv.org/content/10.1101/2020.08.06.20169797v1>
31. Park YJ, Choe YJ, Park O, Park SY, Kim Y-M, Kim J, et al. Early Release - Contact Tracing during Coronavirus Disease Outbreak, South Korea, 2020 - Volume 26, Number 10—October 2020 - Emerging Infectious Diseases journal - CDC. [cited 2020 Aug 1]; Available from: https://wwwnc.cdc.gov/eid/article/26/10/20-1315_article
32. CDC. COVID-19 and Your Health [Internet]. Centers for Disease Control and Prevention. 2020 [cited 2021 Jan 4]. Available from: <https://www.cdc.gov/coronavirus/2019-ncov/if-you-are-sick/quarantine.html>
33. Gallup I. Willingness to Get COVID-19 Vaccine Ticks Up to 63% in U.S. [Internet]. Gallup.com. 2020 [cited 2021 Jan 4]. Available from: <https://news.gallup.com/poll/327425/willingness-covid-vaccine-ticks.aspx>
34. Will M. School Workers May Get Early Shot at COVID-19 Vaccine. Will They Take It? [Internet]. Education Week. 2020 [cited 2021 Jan 4]. Available from: <https://www.edweek.org/leadership/school-workers-may-get-early-shot-at-covid-19-vaccine-will-they-take-it/2020/12>
35. Polack FP, Thomas SJ, Kitchin N, Absalon J, Gurtman A, Lockhart S, et al. Safety and Efficacy of the BNT162b2 mRNA Covid-19 Vaccine. *New England Journal of Medicine* [Internet]. 2020 Dec [cited 2021 Jan 4];383(27):2603–15. Available from: <https://doi.org/10.1056/NEJMoa2034577>
36. Wheaton WD. U.S. Synthetic Population 2010 Version 1.0 Quick Start Guide, RTI International. 2014 May.
37. Digest of Education Statistics, 2019 [Internet]. [cited 2020 Dec 12]. Available from: https://nces.ed.gov/programs/digest/d19/tables/dt19_216.40.asp?current=yes
38. Madewell ZJ, Yang Y, Longini IM, Halloran ME, Dean NE. Household Transmission of SARS-CoV-

2. A Systematic Review and Meta-analysis. JAMA Network Open [Internet]. 2020 Dec [cited 2021 Jan 3];3(12):e2031756. Available from: <https://jamanetwork.com/journals/jamanetworkopen/fullarticle/2774102>
39. Fung HF, Martinez L, Alarid-Escudero F, Salomon JA, Studdert DM, Andrews JR, et al. The Household Secondary Attack Rate of Severe Acute Respiratory Syndrome Coronavirus 2 (SARS-CoV-2): A Rapid Review. Clinical Infectious Diseases [Internet]. 2020 Oct [cited 2021 Jan 3];ciaa1558. Available from: <https://academic.oup.com/cid/advance-article/doi/10.1093/cid/ciaa1558/5921151>
40. Cheng H-Y, Jian S-W, Liu D-P, Ng T-C, Huang W-T, Lin H-H. Contact Tracing Assessment of COVID-19 Transmission Dynamics in Taiwan and Risk at Different Exposure Periods Before and After Symptom Onset. JAMA Internal Medicine [Internet]. 2020 May [cited 2020 Jun 2]; Available from: <https://jamanetwork.com/journals/jamainternalmedicine/fullarticle/2765641>
41. Clapp PW, Sickbert-Bennett EE, Samet JM, Berntsen J, Zeman KL, Anderson DJ, et al. Evaluation of Cloth Masks and Modified Procedure Masks as Personal Protective Equipment for the Public During the COVID-19 Pandemic. JAMA Internal Medicine [Internet]. 2020 Dec [cited 2021 Jan 4]; Available from: <https://jamanetwork.com/journals/jamainternalmedicine/fullarticle/2774266>
42. Chu DK, Akl EA, Duda S, Solo K, Yaacoub S, Schünemann HJ, et al. Physical distancing, face masks, and eye protection to prevent person-to-person transmission of SARS-CoV-2 and COVID-19: A systematic review and meta-analysis. The Lancet [Internet]. 2020 Jun [cited 2021 Jan 4];395(10242):1973–87. Available from: <https://linkinghub.elsevier.com/retrieve/pii/S0140673620311429>
43. Viner RM, Mytton OT, Bonell C, Melendez-Torres GJ, Ward J, Hudson L, et al. Susceptibility to SARS-CoV-2 Infection Among Children and Adolescents Compared With Adults: A Systematic Review and Meta-analysis. JAMA Pediatrics [Internet]. 2020 Sep [cited 2021 Jan 6]; Available from: <https://jamanetwork.com/journals/jamapediatrics/fullarticle/2771181>
44. Dattner I, Goldberg Y, Katriel G, Yaari R, Gal N, Miron Y, et al. The role of children in the spread of COVID-19: Using household data from Bnei Brak, Israel, to estimate the relative susceptibility and infectivity of children [Internet]. Infectious Diseases (except HIV/AIDS); 2020 Jun [cited 2020 Jun 19]. Available from: <http://medrxiv.org/lookup/doi/10.1101/2020.06.03.20121145>
45. Fontanet A, Tondeur L, Madec Y, Grant R, Besombes C, Jolly N, et al. Cluster of COVID-19 in northern France: A retrospective closed cohort study. medRxiv [Internet]. 2020 Apr [cited 2020 Jul

- 9];2020.04.18.20071134. Available from: <https://www.medrxiv.org/content/10.1101/2020.04.18.20071134v1>
46. Byambasuren O, Cardona M, Bell K, Clark J, McLaws M-L, Glasziou P. Estimating the extent of asymptomatic COVID-19 and its potential for community transmission: Systematic review and meta-analysis. *Official Journal of the Association of Medical Microbiology and Infectious Disease Canada* [Internet]. 2020 Dec [cited 2021 Jan 4];5(4):223–34. Available from: <https://jammi.utpjournals.press/doi/10.3138/jammi-2020-0030>
47. Kerr CC, Stuart RM, Mistry D, Abeysuriya RG, Hart G, Rosenfeld K, et al. Covasim: An agent-based model of COVID-19 dynamics and interventions. *medRxiv* [Internet]. 2020 May [cited 2020 Jun 2];2020.05.10.20097469. Available from: <https://www.medrxiv.org/content/10.1101/2020.05.10.20097469v1>
48. Endo A, Centre for the Mathematical Modelling of Infectious Diseases COVID-19 Working Group, Abbott S, Kucharski AJ, Funk S. Estimating the overdispersion in COVID-19 transmission using outbreak sizes outside China. *Wellcome Open Research* [Internet]. 2020 Apr [cited 2020 Jun 11];5:67. Available from: <https://wellcomeopenresearch.org/articles/5-67/v1>
49. Schools and the Path to Zero – Pandemics Explained [Internet]. [cited 2021 Jan 20]. Available from: <https://globalepidemics.org/2020/12/18/schools-and-the-path-to-zero/>
50. CDC. Communities, Schools, Workplaces, & Events [Internet]. Centers for Disease Control and Prevention. 2020 [cited 2020 Nov 25]. Available from: <https://www.cdc.gov/coronavirus/2019-ncov/community/schools-childcare/indicators.html>
51. Link-Gelles R. Limited Secondary Transmission of SARS-CoV-2 in Child Care Programs — Rhode Island, June 1–July 31, 2020. *MMWR Morbidity and Mortality Weekly Report* [Internet]. 2020 [cited 2020 Sep 17];69. Available from: <https://www.cdc.gov/mmwr/volumes/69/wr/mm6934e2.htm>
52. Ladhani S, Ahmad S, Garstang J, Brent AJ, Brent B. Prospective active national surveillance of preschools and primary schools for SARS-CoV-2 infection and transmission in England, June 2020.:23.
53. BWH Press Release - Brigham and Women’s Hospital [Internet]. [cited 2020 Nov 25]. Available from: <https://www.brighamandwomens.org/about-bwh/newsroom/press-releases-detail?id=3684>
54. Althouse BM, Wenger EA, Miller JC, Allard A. Stochasticity and heterogeneity in the transmission dynamics of SARS-CoV-2.:10.
55. Sun K, Viboud C. Impact of contact tracing on SARS-CoV-2 transmission. *The Lancet Infectious Diseases*

[Internet]. 2020 Apr [cited 2020 May 14];0(0). Available from: [https://www.thelancet.com/journals/laninf/article/PIIS1473-3099\(20\)30357-1/abstract](https://www.thelancet.com/journals/laninf/article/PIIS1473-3099(20)30357-1/abstract)

56. Tupper P, Boury H, Yerlanov M, Colijn C. Event-specific interventions to minimize COVID-19 transmission. *Proceedings of the National Academy of Sciences* [Internet]. 2020 Nov [cited 2020 Nov 25]; Available from: <https://www.pnas.org/content/early/2020/11/18/2019324117>

57. Meyerowitz EA, Richterman A, Gandhi RT, Sax PE. Transmission of SARS-CoV-2: A Review of Viral, Host, and Environmental Factors. *Annals of Internal Medicine* [Internet]. 2020 Sep [cited 2020 Nov 25]; Available from: <https://www.acpjournals.org/doi/full/10.7326/M20-5008>

58. Volz E, Mishra S, Chand M, Barrett JC, Johnson R, Geidelberg L, et al. Transmission of SARS-CoV-2 Lineage B.1.1.7 in England: Insights from linking epidemiological and genetic data. *medRxiv* [Internet]. 2021 Jan [cited 2021 Jan 6];2020.12.30.20249034. Available from: <https://www.medrxiv.org/content/10.1101/2020.12.30.20249034v2>

59. Vogel Jan. 15 G, 2021, Pm 4. School risk calculations scrambled by fast-spreading virus strains [Internet]. *Science AAAS*. 2021 [cited 2021 Jan 22]. Available from: <https://www.sciencemag.org/news/2021/01/new-coronavirus-variant-scrambles-school-risk-calculations>

60. Han MS, Choi EH, Chang SH, Jin B-L, Lee EJ, Kim BN, et al. Clinical Characteristics and Viral RNA Detection in Children With Coronavirus Disease 2019 in the Republic of Korea. *JAMA Pediatrics* [Internet]. 2020 Aug [cited 2020 Dec 13]; Available from: <https://doi.org/10.1001/jamapediatrics.2020.3988>

61. Lauer SA, Grantz KH, Bi Q, Jones FK, Zheng Q, Meredith HR, et al. The Incubation Period of Coronavirus Disease 2019 (COVID-19) From Publicly Reported Confirmed Cases: Estimation and Application. *Annals of Internal Medicine* [Internet]. 2020 Mar [cited 2020 Jun 2];172(9):577–82. Available from: <https://www.acpjournals.org/doi/10.7326/M20-0504>

62. Li Q, Guan X, Wu P, Wang X, Zhou L, Tong Y, et al. Early Transmission Dynamics in Wuhan, China, of Novel Coronavirus–Infected Pneumonia. *New England Journal of Medicine* [Internet]. 2020 Jan [cited 2020 Mar 14];0(0):null. Available from: <https://doi.org/10.1056/NEJMoa2001316>

63. Gatto M, Bertuzzo E, Mari L, Miccoli S, Carraro L, Casagrandi R, et al. Spread and dynamics of the COVID-19 epidemic in Italy: Effects of emergency containment measures. *Proceedings of the National Academy of Sciences* [Internet]. 2020 May [cited 2020 Jun 2];117(19):10484–91. Available from: <http://www.pnas.org/lookup/doi/10.1073/pnas.2004978117>

64. He X, Lau EHY, Wu P, Deng X, Wang J, Hao X, et al. Temporal dynamics in viral shedding and transmissibility of COVID-19. *Nature Medicine* [Internet]. 2020 Apr [cited 2020 Apr 29];1–4. Available from: <https://www.nature.com/articles/s41591-020-0869-5>
65. He D, Zhao S, Lin Q, Zhuang Z, Cao P, Wang MH, et al. The relative transmissibility of asymptomatic COVID-19 infections among close contacts. *International Journal of Infectious Diseases* [Internet]. 2020 May [cited 2020 Jun 2];94:145–7. Available from: <http://www.sciencedirect.com/science/article/pii/S1201971220302502>
66. Firth JA, Hellewell J, Klepac P, Kissler SM, CMMID COVID-19 working group, Kucharski AJ, et al. Combining fine-scale social contact data with epidemic modelling reveals interactions between contact tracing, quarantine, testing and physical distancing for controlling COVID-19 [Internet]. *Epidemiology*; 2020 May [cited 2020 Jun 2]. Available from: <http://medrxiv.org/lookup/doi/10.1101/2020.05.26.20113720>
67. Davies NG, Klepac P, Liu Y, Prem K, Jit M, Eggo RM. Age-dependent effects in the transmission and control of COVID-19 epidemics. *Nature Medicine* [Internet]. 2020 Aug [cited 2021 Jan 24];26(8):1205–11. Available from: <https://www.nature.com/articles/s41591-020-0962-9>
68. Zhu Y, Bloxham CJ, Hulme KD, Sinclair JE, Tong ZWM, Steele LE, et al. A meta-analysis on the role of children in SARS-CoV-2 in household transmission clusters. *medRxiv* [Internet]. 2020 Dec [cited 2021 Jan 24];2020.03.26.20044826. Available from: <https://www.medrxiv.org/content/10.1101/2020.03.26.20044826v2>
69. Exclusive: Kids catch and spread coronavirus half as much as adults, Iceland study confirms [Internet]. *Science*. 2020 [cited 2021 Jan 24]. Available from: <https://www.nationalgeographic.com/science/2020/12/we-now-know-how-much-children-spread-coronavirus/>
70. Kim J, Choe YJ, Lee J, Park YJ, Park O, Han MS, et al. Role of children in household transmission of COVID-19. *Archives of Disease in Childhood* [Internet]. 2020 Aug [cited 2021 Jan 24]; Available from: <https://adc.bmj.com/content/early/2020/08/06/archdischild-2020-319910>
71. Lee EJ, Kim DH, Chang SH, Suh SB, Lee J, Lee H, et al. Absence of SARS-CoV-2 Transmission from Children in Isolation to Guardians, South Korea - Volume 27, Number 1—January 2021 - *Emerging Infectious Diseases journal - CDC*. [cited 2021 Jan 24]; Available from: https://wwwnc.cdc.gov/eid/article/27/1/20-3450_article
72. Goldstein E, Lipsitch M, Cevik M. On the Effect of Age on the Transmission of SARS-CoV-2 in Households, Schools, and the Community. *The Journal of Infectious Diseases* [Internet]. 2020 Oct [cited 2021 Jan

24];(jiaa691). Available from: <https://doi.org/10.1093/infdis/jiaa691>

73. Carcione D, Giele CM, Goggin LS, Kwan KS, Smith DW, Dowse GK, et al. Secondary attack rate of pandemic influenza A(H1N1)2009 in Western Australian households, 29 May–7 August 2009. *Eurosurveillance* [Internet]. 2011 Jan [cited 2020 Jun 19];16(3):19765. Available from: <https://www.eurosurveillance.org/content/10.2807/ese.16.03.19765-en>

74. Odaira F, Takahashi H, Toyokawa T, Tsuchihashi Y, Kodama T, Yahata Y, et al. Assessment of secondary attack rate and effectiveness of antiviral prophylaxis among household contacts in an influenza A(H1N1)v outbreak in Kobe, Japan, May–June 2009. *Eurosurveillance* [Internet]. 2009 Sep [cited 2020 Jun 19];14(35):19320. Available from: <https://www.eurosurveillance.org/content/10.2807/ese.14.35.19320-en>

75. Sikora C, Fan S, Golonka R, Sturtevant D, Gratrix J, Lee BE, et al. Transmission of pandemic influenza A (H1N1) 2009 within households: Edmonton, Canada. *Journal of Clinical Virology* [Internet]. 2010 Oct [cited 2020 Jun 19];49(2):90–3. Available from: <http://www.sciencedirect.com/science/article/pii/S1386653210002593>

76. Sugimoto JD, Borse NN, Ta ML, Stockman LJ, Fischer GE, Yang Y, et al. The Effect of Age on Transmission of 2009 Pandemic Influenza A (H1N1) in a Camp and Associated Households. *Epidemiology* (Cambridge, Mass) [Internet]. 2011 Mar [cited 2020 Jun 19];22(2). Available from: <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3755879/>

77. Klick B, Nishiura H, Ng S, Fang VJ, Leung GM, Malik Peiris JS, et al. Transmissibility of seasonal and pandemic influenza in a cohort of households in Hong Kong in 2009. *Epidemiology* (Cambridge, Mass) [Internet]. 2011 Nov [cited 2020 Jun 19];22(6):793–6. Available from: <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3206962/>

78. Wu JT, Ma ESK, Lee CK, Chu DKW, Ho P-L, Shen AL, et al. The Infection Attack Rate and Severity of 2009 Pandemic H1N1 Influenza in Hong Kong. *Clinical Infectious Diseases* [Internet]. 2010 Nov [cited 2020 Jun 19];51(10):1184–91. Available from: <https://academic.oup.com/cid/article/51/10/1184/393386>

79. Bi Q, Wu Y, Mei S, Ye C, Zou X, Zhang Z, et al. Epidemiology and transmission of COVID-19 in 391 cases and 1286 of their close contacts in Shenzhen, China: A retrospective cohort study. *The Lancet Infectious Diseases* [Internet]. 2020 Aug [cited 2021 Jan 24];20(8):911–9. Available from: [https://www.thelancet.com/journals/laninf/article/PIIS1473-3099\(20\)30287-5/abstract](https://www.thelancet.com/journals/laninf/article/PIIS1473-3099(20)30287-5/abstract)

80. Fateh-Moghadam P, Battisti L, Molinaro S, Fontanari S, Dallago G, Binkin N, et al. Contact tracing during

Phase I of the COVID-19 pandemic in the Province of Trento, Italy: Key findings and recommendations. medRxiv [Internet]. 2020 Jul [cited 2021 Jan 24];2020.07.16.20127357. Available from: <https://www.medrxiv.org/content/10.1101/2020.07.16.20127357v1>

81. Li F, Li Y-Y, Liu M-J, Fang L-Q, Dean NE, Wong GWK, et al. Household transmission of SARS-CoV-2 and risk factors for susceptibility and infectivity in Wuhan: A retrospective observational study. The Lancet Infectious Diseases [Internet]. 2021 Jan [cited 2021 Jan 24];0(0). Available from: [https://www.thelancet.com/journals/laninf/article/PIIS1473-3099\(20\)30981-6/abstract](https://www.thelancet.com/journals/laninf/article/PIIS1473-3099(20)30981-6/abstract)

82. Grijalva CG. Transmission of SARS-COV-2 Infections in Households — Tennessee and Wisconsin, April–September 2020. MMWR Morbidity and Mortality Weekly Report [Internet]. 2020 [cited 2021 Jan 24];69. Available from: <https://www.cdc.gov/mmwr/volumes/69/wr/mm6944e1.htm>

83. Lopez Bernal J, Panagiotopoulos N, Byers C, Garcia Vilaplana T, Boddington NL, Zhang X, et al. Transmission dynamics of COVID-19 in household and community settings in the United Kingdom [Internet]. Epidemiology; 2020 Aug [cited 2021 Jan 24]. Available from: <http://medrxiv.org/lookup/doi/10.1101/2020.08.19.20177188>

84. Leclerc QJ, Fuller NM, Knight LE, CMMID COVID-19 Working Group, Funk S, Knight GM. What settings have been linked to SARS-CoV-2 transmission clusters? Wellcome Open Research [Internet]. 2020 Jun [cited 2021 Jan 24];5:83. Available from: <https://wellcomeopenresearch.org/articles/5-83/v2>

85. Jones RD. COVID-19 Trends in Florida K-12 Schools, August 10 – November 14, 2020. medRxiv [Internet]. 2020 Dec [cited 2021 Jan 24];2020.11.30.20241224. Available from: <https://www.medrxiv.org/content/10.1101/2020.11.30.20241224v1>

86. National COVID-19 School Response Dashboard [Internet]. [cited 2021 Jan 24]. Available from: https://statsiq.co1.qualtrics.com/public-dashboard/v0/dashboard/5f78e5d4de521a001036f78e#/dashboard/5f78e5d4de521a001036f78e?pageId=Page_f6071bf7-7db4-4a61-942f-ade4cce464de

87. Covid-19: Sparkletots Preschool cluster increases to 20 cases, forming 3rd largest local cluster [Internet]. [cited 2021 Jan 24]. Available from: <https://mothership.sg/2020/03/covid-19-sparkletots-preschool-coi/>

88. Heavey L, Casey G, Kelly C, Kelly D, McDarby G. No evidence of secondary transmission of COVID-19 from children attending school in Ireland, 2020. Eurosurveillance [Internet]. 2020 May [cited 2021 Jan 24];25(21). Available from: <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7268273/>

89. Oster E. Opinion Schools are not spreading covid-19. This new data makes the case. Washington Post [Internet]. 2020 Nov [cited 2021 Jan 24]; Available from: <https://www.washingtonpost.com/opinions/2020/11/20/covid-19-schools-data-reopening-safety/>
90. COVID Testing Results [Internet]. web. [cited 2021 Jan 24]. Available from: <https://www.schools.nyc.gov/school-year-20-21/return-to-school-2020/health-and-safety/covid-19-testing/covid-testing-results>
91. Second school in Thorncliffe neighbourhood closed over COVID; third York Catholic school shuts down [Internet]. thestar.com. 2020 [cited 2021 Jan 24]. Available from: <https://www.thestar.com/news/gta/2020/12/07/tdsb-closes-thorncliffe-park-elementary-school-amid-rising-number-of-students-testing-positive-for-covid-19.html>
92. Monod M, Blenkinsop A, Xi X, Hebert D, Bershan S, Bradley VC, et al. Age groups that sustain resurging COVID-19 epidemics in the United States. medRxiv [Internet]. 2020 Sep [cited 2021 Jan 24];2020.09.18.20197376. Available from: <https://www.medrxiv.org/content/10.1101/2020.09.18.20197376v1>
93. Harris D, Ziedan E, Hassig S. The Effects of School Reopenings on COVID-19 Hospitalizations [Internet]. [cited 2021 Jan 24]. Available from: <https://www.reachcentered.org/publications/the-effects-of-school-reopenings-on-covid-19-hospitalizations>