Revisiting early-stage COVID-19 strategy options [version 1; peer review: awaiting peer review]

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Abstract

**Background:** Early-stage interventions in a potential pandemic are important to understand as they can make the difference between runaway exponential growth that is hard to turn back and stopping the spread before it gets that far. COVID19 is an interesting case study because there have been very different outcomes in different localities. These variations are best studied after the fact if precision is the goal; while a pandemic is still unfolding less precise analysis is of value in attempting to guide localities in the early stages to learn lessons of those that preceded them.

**Methods:** I examine three factors that could differentiate strategy: asymptomatic spread, differences in use of the Bacillus Calmette-Guerin (BCG) tuberculosis vaccine and cloth face masks. **Results:** These differences point to the possibility of alternative strategies to prevent COVID-19 from entering the runaway phase. The most promising is testing all contacts of anyone who has tested positive, not only those who are symptomatic. In principle such a study should be able to demonstrate not only what fraction are asymptomatic but also measure asymptomatic transmission by another round of contact tracing. Should this demonstrate asymptomatic transmission, then all contacts of anyone who tests positive should be isolated and tested, and only released from quarantine when it is clear that they are past incubation and not positive.

**Conclusions:** Before a COVID-19 outbreak reaches the runaway phase where exponential growth makes accurate trace-and-test impossible, a strategy that takes into account asymptomatic transmission is viable and should be attempted to avoid the necessity for a prolonged lockdown.

**Keywords**
COVID-19, early-stage pandemic intervention, asymptomatic transmission
Introduction

As the COVID-19 pandemic has spread, its outcomes have differed by locality. In some, it has been contained quickly. In others, the rapid growth has slowed but not stopped. In many, the rapid growth has driven health systems to the point of collapse. New York state is the epicenter of the United States epidemic. What adds urgency to the search for alternative containment strategies is the fact that mortality rate (deaths as a fraction of population) in New York state on 22 April 2020 surpassed 1000 per million. If scaled to the entire country, this would be over 300,000 deaths\(^1\).

Since the pandemic is still playing out it is useful to reflect on the positive and negative outcomes and to try to map a way ahead for localities where it has not gone past the stage where it can easily be contained. Given that no known remedy exists for the disease, and that it is spreading too fast to rely on a vaccine to avoid major health or economic problems, non-pharmaceutical interventions (NPI) are the most critical thing to get right.

A number of NPIs have been tried from social distancing to complete lockdowns. The consequence of acting too slowly is the risk of crashing a health system, which has hit some of the best in the world (Italy for example has historically been near the top of the World Health Organization’s ranking\(^2\); and the epicentre of the COVID-19 outbreak is in the north\(^3\), which has Italy’s strongest health resources\(^4\). Though the pandemic at time of writing is still developing fast, it is worth reviewing options for countries able to avoid runaway exponential growth.

By way of example, I look at options for South Africa, which embarked on a 21-day lock down\(^1\) that started at about the time when 1,000 cases were reported (midnight, 26 March 2020; subsequently extended by another two weeks to 30 April).

I examine case studies in other localities encompassing the variability in outcomes and assess likely contributing factors to this variance. Given the shortness of time to make decisions, I do not attempt to develop a rigorous model but rely on extracting meaning from these case studies.

In my opinion, ignoring asymptomatic spread is a major error; a relatively simple experiment in a country like South Africa at the early stage of spread could validate this opinion. If proved correct, many lives could be saved. If proved incorrect, the cost is low relative to the benefit. I therefore urge that the experiment be carried out as a matter of urgency.

In the remainder of the paper, I describe significant unknowns, work through case studies and examine other factors, leading to the conclusion that a study of asymptomatic spread is the most urgent gap in current knowledge that must be addressed to assess options for containment.

Significant unknowns

Because testing started under pressure, standards are not consistent\(^5\). That means statistics like case fatality rate are problematic to compare across localities.

Another big unknown is the true number of cases since many that were not serious enough to require hospitalization may have resolved without being counted where testing was inadequate; if the asymptomatic fraction is as high as claimed in some instances, that also skews the case fatality rate high.

All of these factors also result in difficulty in establishing an accurate value for the basic reproduction rate \(R_0\), the mean number of infections per infected case. The value of \(R_0\) matters for computing the herd immunity level, widely reported in the mass media as 60% of the population, the number used to justify the initial British response of allowing it to run through the population\(^6\).

Herd immunity occurs when the fraction of the population immunised (either by vaccination or by acquiring immunity post-infection) exceeds the threshold \(P_{\text{herd}}\) in Equation 1,

\[
P_{\text{herd}} = \frac{1}{E} \left(1 - \frac{1}{R_0}\right)
\]

If immunisation is 100% effective \((E = 1)\), Equation 1 becomes\(^6\):

\[
P_{\text{herd}} = 1 - \frac{1}{R_0}
\]

Figure 1 illustrates how \(P_{\text{herd}}\) varies with \(R_0\) (assuming \(E = 1\); in the absence of a vaccine, this means that any recovered cases cannot be reinfected). For seasonal influenza, if \(R_0 = 1.3\), the herd immunity threshold is 23%. For \(R_0 = 2.5\), the herd immunity threshold is 60%. However, if the true COVID-19 \(R_0\) value is significantly higher, so is the the herd immunity threshold. For example, if \(R_0 = 4\), \(P_{\text{herd}} = 75\%\).

Even for influenza, \(R_0\) can vary widely depending on the strain. For the H1N1 strain, \(R_0\) was estimated at 1.4–1.6; for the 1918

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\(^1\)The data source for cases and fatalities, unless otherwise stated, is the Worldometers web site.

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Figure 1. Herd immunity varying with basic reproduction rate \(R_0\), if \(R_0\) for COVID-19 is 2.5, herd immunity occurs after 80% of the population is infected (green arrow). The mean \(R_0\) value for seasonal influenza results in herd immunity at about 23% of the population (red arrow).
flu, the estimated $R_0$ range is 1.4–2.8 and even seasonal flu has a wide $R_0$ range of 0.9–2.1\(^8\).

One study reports COVID-19 $R_0$ values varying from 1.4 to 3.8\(^9\). Another narrows the range to 2.24 to 3.58\(^10\). A model with $R_0 = 2.68$ yields a doubling time of 6.4 days\(^11\). The doubling time was far shorter than this during peak growth in places where it was not under control. In the United States, for example, doubling time was less than 3 days 1–24 March 2020 (see Figure 2, based on the rule-of-70\(^12\)). That rapidity of growth suggests that $R_0$ is on the high rather than low side of , but the low level of testing in the USA makes it difficult to derive robust measures.

With so much uncertainty, relying on herd immunity is folly, as was discovered in the UK\(^6\).

Broadly speaking, countries with a tradition of personal liberty and rejecting authority have found it hard to adapt to NPIs like social distancing\(^13\). Informal settlements and other high-density dwellings of the poor also make social distancing hard\(^14\). For this reason, it is important to explore all alternatives including those that were missed in early stages so that other countries that are being hit later can learn the right lessons.

**Methods**

I use Python 3.7.4 within the Jupyter Notebook environment version 6.0.3 to do any analysis. All the code used is archived on GitHub\(^15\) and is labelled as version 1.1.1.

Data is from the Worldometers web site and papers cited; data used in analysis is embedded in the GitHub archive\(^16\). A snapshot of the Worldometers site containing the data used is archived at Webarchive.

Herd immunity is calculated based on the standard formula assuming exponential growth using Equation 2.

I calculate doubling time for March 2020 using the rule-of-70, which assumes exponential growth and is accurate under that assumption:

$$t_{double} = \frac{70}{r_{\text{growth}}}$$  \hspace{1cm} (3)

Where the growth rate is expressed as a fraction of 100. For example, if the growth rate $r_{\text{growth}} = 40\%$ then:

$$t_{double} = \frac{70}{40} = 1.75$$ \hspace{1cm} (4)

The number 70 derives from the fact that $0.7 \approx \ln(2)$\(^12\).

**Case studies**

I consider three cases: instances where the disease has run its course, examples where containment has slowed the growth to the point of being manageable and finally examples where growth has severely stressed health systems.

These examples are not meant to be exhaustive but provide archetypes of different types of outcome.

**Run to completion**

There are two types of localities where the disease has effectively run to completion: cruise liners and a village in Italy, Vo’ Euganeo.

I take Diamond Princess, a widely cited case, as an example of the first kind.

**Cruise liner**

Diamond Princess is one of the most documented examples. Of the 3,711 passengers and crew on board, 634 tested positive out of the 3,063 tested and 328 were asymptomatic\(^16\) – meaning that over 50% were asymptomatic.

What is startling about these numbers is how few tested positive. Given the constrained environment of a cruise liner, a higher rate of social mixing would be expected than in normal living conditions.

However the major figure to take way from this is the fact that over half were asymptomatic.

**Italian village**

For the Italian village, the number of cases that tested positive but were asymptomatic were in the range of 50–75%\(^17\).

What is particularly noteworthy is that 100% of the population was tested so no asymptomatic cases were missed. This completeness of coverage does not guarantee that a similar result would occur in a different population but at least provides a clear data point. All positive cases were isolated. The number who were ill, fell from 88 to 7 in 10 days.

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\(^2\)The wide range is not explained – possibly defining asymptomatic is imprecise.
By contrast with this strategy, on 25 February 2020, the strategy for the rest of Italy changed from broad testing of all contacts to a focus on only testing those who had clear symptoms and required hospitalization. This decision was taken when the country had less than 100 cases.

It is a puzzle why it should be more important to test cases who must be hospitalized anyway rather than find those in the community who may be contagious.

I quote the advice of Italian immunologist Prof Romagnani (via Google Translate):

> It is therefore absolutely essential to extend the swabs to the majority of the population, in particular to risk categories (i.e. exposed to multiple contacts), and therefore isolate the virus positive subjects and their contacts, even if asymptomatic, as early as possible. In particular, it is absolutely necessary to swab all those who have a high probability of transmitting the virus, especially if they live in closed communities with multiple and close contacts. Finally, it is very important that all those at risk wear masks, not so much to protect themselves from infection, but rather to protect others from themselves, even when they do not show symptoms.

Even if we take the lower figure of 50% of the Vo’Euganeo being asymptomatic, ignoring this factor in testing is a major error. However, in a larger population it is not practical to test everyone and not necessarily useful as a person tested negative could subsequently be infected. Instead, treating everyone exposed to a known case as infected until proved otherwise would catch a significant fraction of the asymptomatic cases.

**Slower growth**

Several countries have managed to contain exponential growth after a major outbreak. Two examples are China and South Korea.

Since China had the first outbreak, they did not act fast enough and had to close down a major part of their economy. Much of Hubei province was effectively placed under quarantine on 23 January 2020, using methods such as tracking social media to enforce it. Since China is a large country, it was possible to isolate the infected region and rush in resources from the rest of the country. Overall the measures used seem unlikely to be transferable to most other countries.

South Korea instead took an approach of rapid and comprehensive testing while isolating all known cases and all their contacts. With no lockdown, they brought the increase off the exponential trend. That supports my opinion of how the Italian test strategy went wrong.

**Rapid growth**

The United States only showed signs of breaking exponential growth after over 300,000 cases were reported on 4 April 2020. Some days during March 2020, growth was at 40% or more per day with a doubling time of two days and, up to 24 March, doubling time was generally three days or less.

Over the period 2–20 March 2020, growth in the US ranged from 24% to 49% per day. Figure 3 illustrates how the United States had no option but to implement increasingly country-wide lockdowns. Other countries that resisted this strategy ran into exactly the same issue: exponential growth defeated their health systems.

In a relatively large country like the US, if growth is not even across the country, there is the option to rush resources to the hotspots as was done in China. However, this option is not as attractive as stopping the spread much earlier as the hotspots become extremely resource intensive. New York state for example has been widely reported as estimating a requirement of 40,000 ventilators at peak.

Localities that have suffered this sort of resource intensity have also generally had higher case fatality rates. This could in part be an artefact of inadequate testing that under-counts cases. However Italy (13% case fatality) has a higher per capita rate of testing than Germany (3.4% case fatality) as of 20 April 2020: 23,985 per million population vs. 20,629 for Germany. Under-counting asymptomatic cases is not likely to be the sole issue.

**Other factors**

An important factor to consider is comorbidities. South Africa has high rates of tuberculosis and HIV infection, both of which are significant risk factors for any pulmonary disease. In one study in Italy, out of 355 patients who died, only 3 (0.8%) had no prior condition. 25.1% had one condition, 25.6% had two and 48.5% had three or more.

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**Figure 3. USA worst-case scenarios without lockdown.** The higher (red) curve represents 40% growth per day from the reported cases of 20 March 2020, totalling 19,551. The lower (blue) curve represents growth of 24% per day from the same starting point.

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1 For example: [https://abc7ny.com/ventilators-coronavirus-new-york-ny-cases-in/6086097/](https://abc7ny.com/ventilators-coronavirus-new-york-ny-cases-in/6086097/).
Balanced against this is statistical evidence that differences in national coverage of the Bacillus Calmette-Guerin (BCG) tuberculosis vaccine explain differences in case fatality rates\textsuperscript{27}. This study is not peer reviewed and therefore should not be relied on too strongly. A pure statistical study without direct causal evidence points to the need to establish causality rather than signifying causality. However, there would be merit in identifying any cohorts who missed childhood BCG vaccination and provide them with this additional extra protection, provided that it is not contra-indicated. BCG is already administered on a mass scale and its risks are well understood.

Study of case outcomes in a country like South Africa where BCG is commonly administered but where there is known to be variations in geographic coverage\textsuperscript{25} could help to validate causality.

Weighed against relying on BCG coverage is that Iceland, which does not have mandatory BCG vaccination\textsuperscript{24}, was very successful in slowing the spread by an aggressive testing programme, including quarantining everyone who had contacted a known case\textsuperscript{25}.

Another possible mitigating factor is that existing HIV remedies are among those being investigated for efficacy against COVID-19\textsuperscript{26}. However, it would be foolhardy to rely on this as a mitigating factor without solid evidence that it applies widely enough to matter.

Finally there is indirect evidence of asymptomatic spread in the apparent efficacy of masks in slowing spread. While early WHO advice was against the asymptomatic mask wearing, this was part of an advisory that aimed to prevent a run on medical-grade masks\textsuperscript{37}. More recently there is support for cloth masks being worn by the public\textsuperscript{28} and experimental evidence that surgical masks block aerosol transmission of coronaviruses and influenza viruses\textsuperscript{39}. While some still doubt the evidence, the fact that mandatory mask wearing was a factor in reducing the spread of the 1918 influenza pandemic\textsuperscript{30} supports the case for encouraging mask-wearing as a protection for the COVID-19 pandemic provided that this does not deplete supplies of medical-grade masks and there is a programme of public education on use of masks to avoid self-infection.

Given the high risk associated with comorbidities, it is premature to place too much reliance on any mitigating factors other than slowing transmission. NPIs are the main game until pharmaceutical options become viable on a large scale.

Conclusions
The most significant finding out of this review is that we do not know enough about asymptomatic transmission. This is a problem easily remedied in early-stage spread.

A very simple thing would be to test all contacts of as many positive cases as possible and check how many were missed by not testing the asymptomatic. This approach would give a clearer picture of the potential for asymptomatic spread; any who test positive out of this cohort can be followed up by contact-tracing and testing. If a further round of contact tracing finds more positive cases that have been in contact with only asymptomatic cases, that would provide a measure of asymptomatic contagion.

If the experiment to measure asymptomatic spread shows that it is a significant factor, that signifies a change in testing strategy. Should any asymptomatic positives be discovered, more aggressive action should follow: they and all their contacts should be quarantined and released when they no longer test positive, taking into account time for incubation.

South Africa at time of writing has less than 3500 cases; going back to test all contacts is doable on this scale. While it is possible that the data cited in this paper on the asymptomatic fraction is wrong, the cost of finding this out is very low compared with the cost of not containing the spread. As previously noted, Iceland does not do widespread BCG vaccination\textsuperscript{24} and yet has been successful in containing the spread. The Iceland experience indicates that there is no need to rely on questionable evidence that the BCG vaccine provides protection: act early and catch all cases including the asymptomatic and the spread can be curtailed\textsuperscript{25}.

The cost of taking the step I advocate is far lower than the cost of allowing asymptomatic spread to get out of control. If it leads to an effective containment strategy, it will also remove the need for an economically damaging extended lockdown.

**Data availability**

**Source data**
COVID-19 data used was gathered from Worldometer: https://www.worldometers.info/.


**Extended data**
The Python code used to graph herd immunity versus $R_0$ as well graphing doubling time and the US exponential growth scenarios is available from GitHub: https://github.com/philip-mach/herd-immunity.


Data is available under the terms of the BSD 2-clause license.
References


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