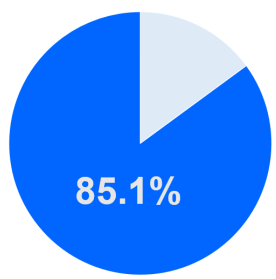
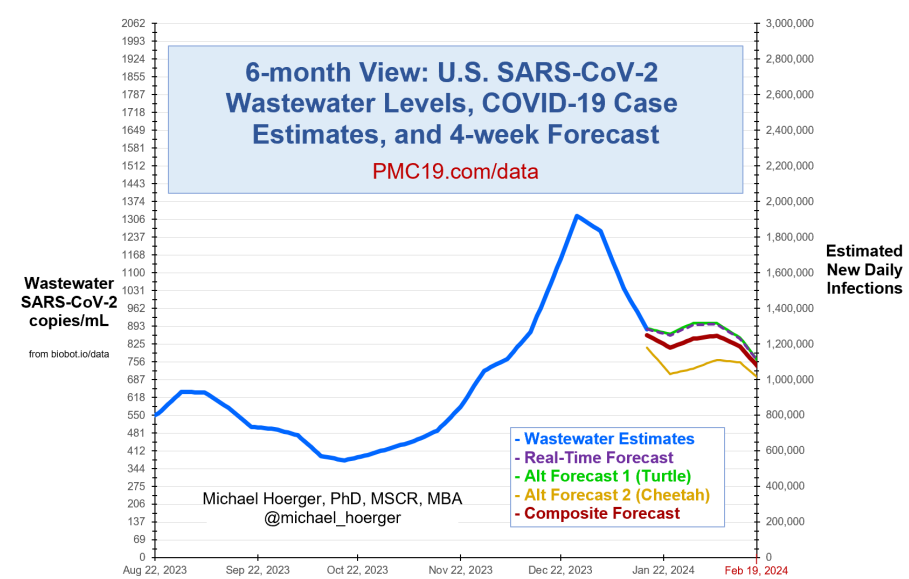
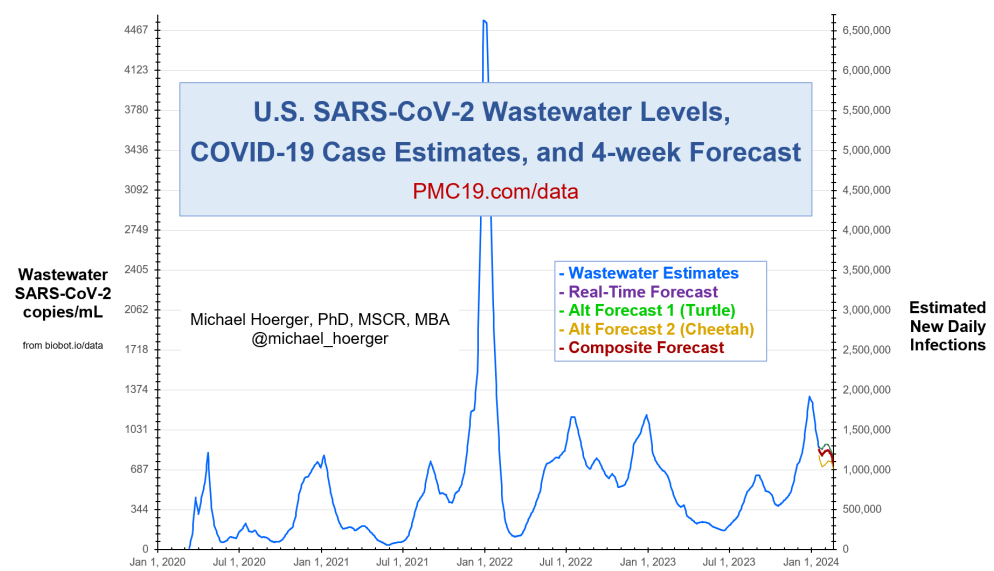


U.S. SARS-CoV-2 Wastewater Levels, COVID-19 Case Estimates, and 4-Week Forecast: Report for January 22, 2024, pmc19.com/data

Michael Hoerger, PhD, MSCR, MBA, Pandemic Mitigation Collaborative



There is more COVID-19 transmission today than during 85.1% of the pandemic.

| CURRENT ESTIMATES FOR January 22, 2024 | |
|--|------------------------|
| Wastewater Levels (copies/mL) | 824 |
| New Daily Cases | 1,200,000 |
| % of Population Infectious | 2.51% (1 in 40 people) |
| New Daily Long COVID Cases | 60,000 to 240,000 |

| WEEKLY ESTIMATES FOR January 22, 2024 | |
|---------------------------------------|----------------------|
| New Weekly Cases | 8,400,000 |
| New Weekly Long COVID Cases | 420,000 to 1,680,000 |

| 2024 CUMULATIVE ESTIMATES AS OF January 22, 2024 | |
|--|------------------------|
| Total 2024 Cases To Date | 32,876,330 |
| Total 2024 Long COVID Cases To Date | 1,644,000 to 6,575,000 |

| 4-WEEK FORECAST FOR February 19, 2024 | |
|---------------------------------------|------------------------|
| Wastewater Levels (copies/mL) | 743 (-10% lower) |
| New Daily Cases | 1,081,000 |
| % of Population Infectious | 2.26% (1 in 44 people) |
| New Daily Long COVID Cases | 54,000 to 216,000 |

| How Does Risk Increase with More Social Contacts? | | | |
|---|------------------------------|------------------|------------------------------|
| Number of People | Chances Anyone is Infectious | Number of People | Chances Anyone is Infectious |
| 1 | 2.5% | 25 | 47.0% |
| 2 | 5.0% | 30 | 53.4% |
| 3 | 7.3% | 35 | 58.9% |
| 4 | 9.7% | 40 | 63.8% |
| 5 | 11.9% | 50 | 72.0% |
| 6 | 14.1% | 75 | 85.1% |
| 7 | 16.3% | 100 | 92.1% |
| 8 | 18.4% | 150 | 97.8% |
| 9 | 20.5% | 200 | 99.4% |
| 10 | 22.5% | 300 | >99.9% |
| 15 | 31.7% | 400 | >99.9% |
| 20 | 39.9% | 500 | >99.9% |

Cite as: Hoerger, M. (2024, January 22). U.S. SARS-CoV-2 wastewater levels, COVID-19 case estimates, and 4-week forecast: Report for January 22, 2024. Pandemic Mitigation Collaborative. <http://www.pmc19.com/data>

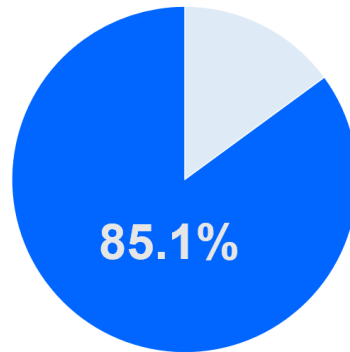
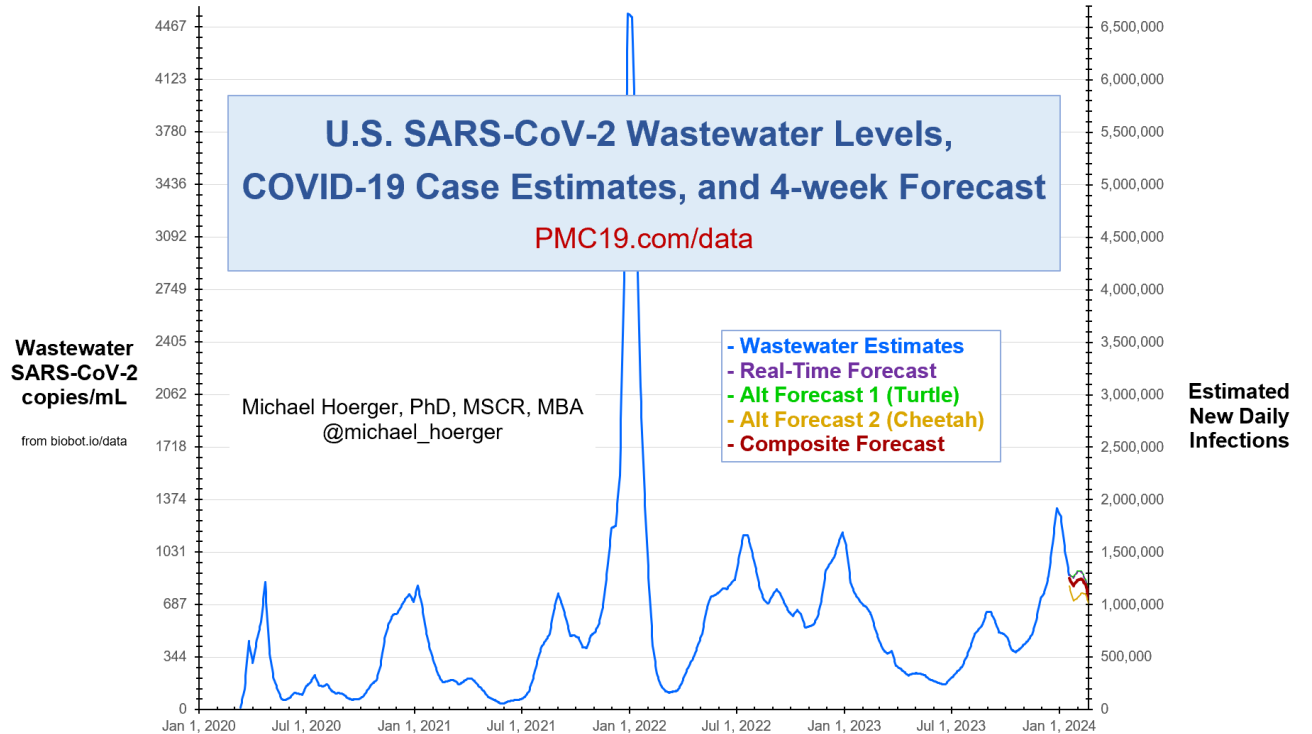
Overview of the Current State of the U.S. COVID-19 Pandemic

We are in the 8th U.S. COVID wave and 2nd biggest all-time. The magnitude of the current wave indicates that the pandemic is not “over.” Indeed, similar or larger surges may occur in the future.

U.S. wastewater levels indicate that COVID transmission is higher than during 85% of the days of the pandemic and lower than during 15% of the days of the pandemic.

- ◆ 2.5% (1 in 40) are infectious
- ◆ 1.2 million COVID cases/day
- ◆ >60,000 Long Covid cases/day

Weekly new cases stand at 8 million. Weekly Long COVID cases resulting from these infections stand at >400,000.



There is more COVID-19 transmission today than during 85.1% of the pandemic.

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|--|------------------------|
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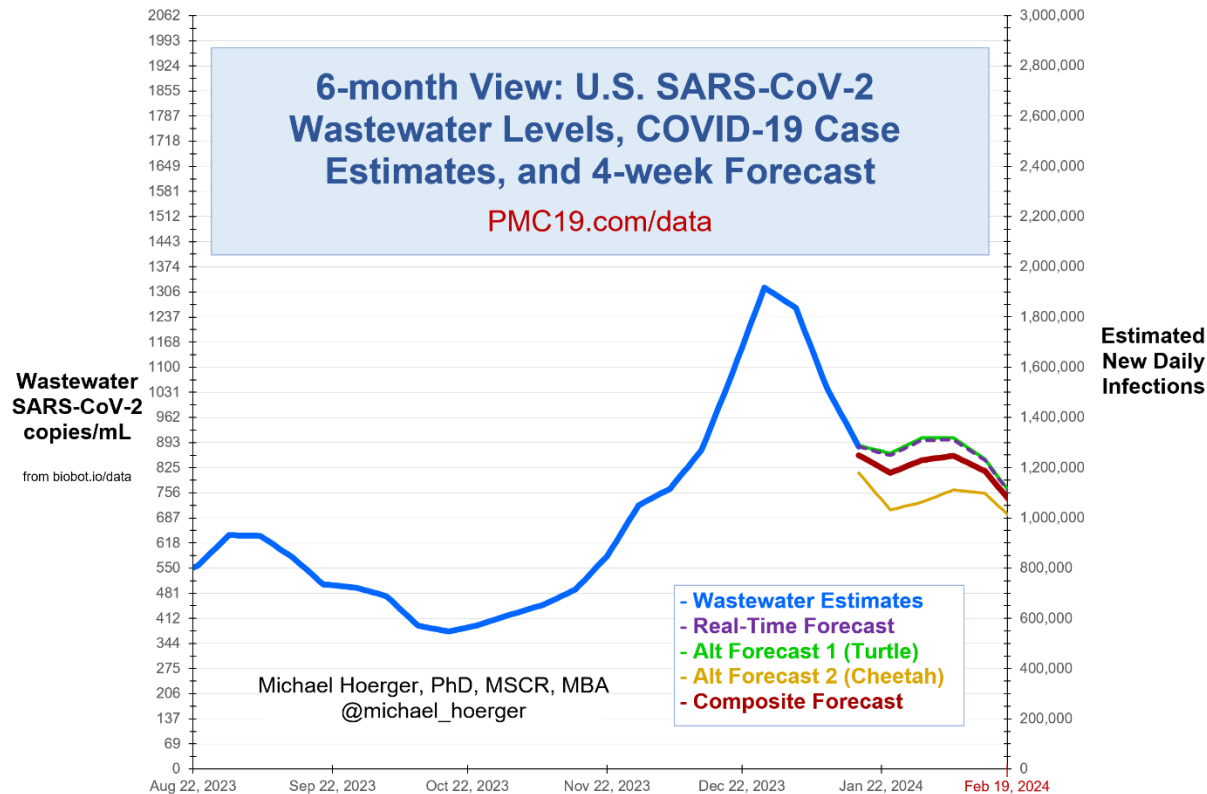
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|--|------------------------|
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| Total 2024 Long COVID Cases To Date | 1,644,000 to 6,575,000 |

Forecast for the Next Month

We are in the 2nd largest U.S. surge of the pandemic. Over the next month, we should see 0.8-1.3 million infections per day. Biobot has made considerable post-hoc corrections to each of the prior 3 weeks of data. This essentially moved the peak back from early January to late December, and at present suggests a slightly more precipitous decline the past few weeks than noted in real time. The details of the forecasting models can be found in the Technical Notes section. Reviewing the graph on the previous page (whole pandemic), you'll notice that after about half the waves, instead of a straightforward decline, there is occasionally a temporary deceleration, or even a small hill. Currently, this is what the forecast is showing. I suspect that the hill in the graphical depiction below is an artifact and we'll head toward a more steady decline, reaching 0.8-0.9 million cases/day in a month. Why? First, the models are based on all data, not just wave-based data, and in my experience tend to be less precise in the very middle of the ascent or descent of a wave. In the 2.0 version of the forecast coming out in a few months, we will likely add complexity to the model to refine the wave phases. Second, Biobot has made huge retroactive corrections to their real-time levels in each recent report. Their current real-time level may be off, and that can throw off the models. Overall, consider that the decline in daily cases may slow, or there might be a small hill on the way, but if lucky, transmission will continue to decline a little faster than depicted.

The CDC has recently approved an updated booster, available to anyone in the U.S. older than 6 months. It is widely available for adults, and now increasingly available for children. Much of the winter surge lies ahead. Use the PMC data and local vaccination location information to help people get vaccinated.



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|--|------------------------|
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| Wastewater Levels (copies/mL) | 743 (-10% lower) |
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| New Daily Long COVID Cases | 54,000 to 216,000 |

Current Risk Based on Number of Social Contacts

This figure shows the chance anyone would be infectious in a group based on group size. For example, in a group of 25-30, there's a 50% chance at least one person is infectious. In a group of 100, there's a >90% chance at least one person is infectious.

| How Does Risk Increase with More Social Contacts? | | | |
|--|-------------------------------------|-------------------------|-------------------------------------|
| Number of People | Chances Anyone is Infectious | Number of People | Chances Anyone is Infectious |
| 1 | 2.5% | 25 | 47.0% |
| 2 | 5.0% | 30 | 53.4% |
| 3 | 7.3% | 35 | 58.9% |
| 4 | 9.7% | 40 | 63.8% |
| 5 | 11.9% | 50 | 72.0% |
| 6 | 14.1% | 75 | 85.1% |
| 7 | 16.3% | 100 | 92.1% |
| 8 | 18.4% | 150 | 97.8% |
| 9 | 20.5% | 200 | 99.4% |
| 10 | 22.5% | 300 | >99.9% |
| 15 | 31.7% | 400 | >99.9% |
| 20 | 39.9% | 500 | >99.9% |

Trustworthy Health Evidence

Transparency

The pandemic has reminded us that trust is essential in public health, and we have made every effort to provide transparency. Every PMC report remains posted publicly. A methodology video describes the gist of the analytic approaches used, with enough detail that others can construct similar models. Dr. Hoerger's biography and publication record are publicly available. Many of those making similar dashboards and forecasts work in private industry or wish to avoid scrutiny so do not post information publicly or do so using anonymous accounts or pseudonyms.

Lack of Conflicts of Interest

These reports are written by Dr. Hoerger who has no financial conflicts of interest. This means no ownership of equity stakes or individual stocks. Dr. Hoerger actively avoids funding from the pharmaceutical industry and other private corporations. He has no investment in mask, HEPA, or other mitigation companies. As a scientist, he is required to complete financial forms as a part of being a university professor as well as when publishing at top medical journals, like JAMA Oncology. An underappreciated aspect of financial conflicts of interest is that clinicians and scientists sometimes gain substantial revenue from social media advertisements (e.g., YouTube, Twitter) and newsletter subscriptions (e.g., Substack, Patreon); this can lead to significance "audience capture" in which the scientist gains increasing financial remuneration for telling an audience what they want to hear, and this is a huge problem among those posting disinformation online. Dr. Hoerger intentionally gains no revenue through this website, social media, nor other platforms, as doing so would undermine his scientific credibility. The closest PMC collaborators are the co-authors and co-investigators noted in the various projects and publications on the PMC homepage, and there are no noteworthy conflicts of interest there either.

Documented Evidence of Accuracy

The PMC Forecasting model can be conceptualized as "participatory action research," meaning that it is public health activism first, iterates and improves with feedback, and is research second. An alternative approach would be to publish a model first and then implement, but that would be a horribly unsuccessful strategy. Using historical data, our forecasting model explains 96% of the variance in 1-week forecasts, which is exceptional. However, implementing forecasts in real-time adds new challenges. Wastewater levels are reported with inconsistent cadence and often corrected retroactively, so forecasting in real-time is more challenging, meaning that real-world estimates of accuracy are more important than those derived from a historical database. Moreover, it has been a learning experience understanding what metrics matter most to the public, and these metrics vary across different points of the pandemic. Accordingly, all reports are public so that users can evaluate accuracy themselves in real-time. However, a scientific report will ultimately document accuracy. Accuracy levels will guide a 2.0 model of the forecast likely to launch in spring or summer of 2024, which may include the same or different alternative forecasts, and add confidence intervals based on historic levels of accuracy.

A useful way to evaluate accuracy is to take one of our public reports and then compare it with what unfolds four weeks later. We believe you will be impressed with the results and that any discrepancies are minor and not of public health significance, given that much of the public is magnitudes off in their subjective estimates of transmission. As an example, in our December 25 report, we noted that our first-posted Christmas Day risk table

(posted October 30, 2023, with that report cited in JAMA Oncology) was quite accurate in characterizing the eventual risk on Christmas Day (noted in the December 25, 2023 report, though Biobot may retroactively adjust wastewater levels marginally).

Posted on December 25

In the U.S., What's the COVID Risk for Christmas Day?

| Number of People | Chances Anyone is Infectious | Number of People | Chances Anyone is Infectious |
|------------------|------------------------------|------------------|------------------------------|
| 1 | 3.5% | 25 | 58.6% |
| 2 | 6.8% | 30 | 65.3% |
| 3 | 10.0% | 35 | 70.9% |
| 4 | 13.2% | 40 | 75.6% |
| 5 | 16.2% | 50 | 82.8% |
| 6 | 19.1% | 75 | 92.9% |
| 7 | 21.9% | 100 | 97.1% |
| 8 | 24.6% | 150 | 99.5% |
| 9 | 27.2% | 200 | 99.9% |
| 10 | 29.7% | 300 | >99.9% |
| 15 | 41.1% | 400 | >99.9% |
| 20 | 50.6% | 500 | >99.9% |

@michael_hoenger Estimated December 25, 2023

Posted on October 30

In the U.S., What's the COVID Risk for Christmas Day?

| Number of People | Chances Anyone is Infectious | Number of People | Chances Anyone is Infectious |
|------------------|------------------------------|------------------|------------------------------|
| 1 | 3.1% | 25 | 54.8% |
| 2 | 6.2% | 30 | 61.5% |
| 3 | 9.1% | 35 | 67.1% |
| 4 | 11.9% | 40 | 72.0% |
| 5 | 14.7% | 50 | 79.6% |
| 6 | 17.4% | 75 | 90.8% |
| 7 | 19.9% | 100 | 95.8% |
| 8 | 22.5% | 150 | 99.1% |
| 9 | 24.9% | 200 | 99.8% |
| 10 | 27.2% | 300 | >99.9% |
| 15 | 37.9% | 400 | >99.9% |
| 20 | 47.0% | 500 | >99.9% |

@michael_hoenger Estimated October 30, 2023

Negligible differences between the long-range forecast and today's table. In a group of 20, for example, the forecasted risk was 47%, and today we estimate it at 50.6%. Hopefully this helped with planning.

Documented Evidence of Validity Relative to Other Data Sources

The public has genuine reasons to ask questions about wastewater-derived case estimates and forecasts. These concerns fall into two main themes. One, much of the general public has observed how important metrics (reported case counts, positivity ratio) have gone from valuable to fallible as testing has declined and largely switched from reported PCR tests to unreported home rapid tests. Two, with each new subvariant, some also wonder whether wastewater data could over or underestimate transmission. With regard to the fallibility of testing-based data, a tremendous benefit of wastewater is that it does not rely on self-report. With regard to subvariants, it is important to know that the process of converting raw

wastewater into useful, normalized metrics of viral levels is overseen by wastewater scientists with specialized expertise in addressing our most obvious concerns (e.g., new subvariants, regional differences in demographics, changes in rain water). It is a good example of the Dunning-Kruger Effect for anyone to assume they have more expertise than the scientific experts who translate sewage into viral level estimates, but our public health infrastructure fosters much skepticism. Moreover, by outsourcing wastewater measurement to third parties who must compete with one another, methodology surrounding wastewater quantification remains proprietary to protect competitive interests, and this reduces transparency and trust. Fortunately, there are two easy checks on wastewater validity.

Validity Check #1: Independent Cross-Cultural Comparisons. If the PMC model provides a useful estimate of cases, it should converge with other independent estimates. Unfortunately, other U.S. based models are private or anonymous, so not appropriate comparisons. Nonetheless, there are high-quality estimates outside the U.S. For example, Dr. Moriarty's lab uses wastewater data in Canada to make similar estimates, and the U.K. runs a strong testing surveillance program to assess case rates. Despite being from independent groups, using slightly to considerably different methodology, and in different countries, all of the estimates are within the same ballpark. The U.S. and U.K. data have tracked very closely. The current estimates in Canada are almost identical to those for the Northeastern U.S. at its peak.

International COVID Statistics from High-Quality Data Sources

Compiled by the Pandemic Mitigation Collaborative (PMC19.com), January 23, 2024

| | U.S.A. | Canada | U.K. |
|--|-----------------------|-----------------------|-----------------------|
| % of Population Actively Infectious | 2.5% (1 in 40) | 7.7% (1 in 13) | 2.3% (1 in 43) |
| Chances Someone is Infectious... | | | |
| In a Group of 10 | 22.5% | 55.1% | 20.8% |
| In a Group of 30 | 53.4% | 90.9% | 50.2% |
| In a Group of 50 | 72.0% | 98.2% | 68.8% |

Primary Data Source
Reference

Wastewater
Michael Hoerger, PhD, MSCR, MBA
@michael_hoerger
pmc19.com/data

Wastewater
Tara Moriarty, PhD
@MoriartyLab
covid19resources.ca

Surveillance Testing
Alex Glaser, PhD
UK Health Service Agency
tinyurl.com/pmckukhsa

Last Updated

January 23, 2024

January 21, 2024

January 10, 2024

Validity Check #2: Examining Consistency Relative to Other Metrics. For the past 18+ months, the relationship between self-reported cases and estimates of true cases has remained remarkably consistent across various indicators of each variable. A simple heuristic is that there are 25 actual cases for every 1 self-reported case tracked in federal statistics. This fluctuates a bit depending on whether in a lull, rising surge, or falling surge, often in the 15-30x range. Simply review IHME self-reported versus actual case data, available through April 1, 2023 (<https://covid19.healthdata.org/united-states-of-america?view=infections-testing&tab=trend&test=infections>). The same is true for our wastewater-derived estimates of true cases, relative to self-reports of actual cases provided by states and aggregated nationally by BNO News (<https://twitter.com/BNONews>) and others. If this relationship were to decouple suddenly, such that estimates of true cases were 100x or 1,000x that of self-reported cases, that would suggest that a particular subvariant may be disrupting wastewater estimates in ways that were outsmarting wastewater scientists, but there is no evidence for such a claim at present.

PMC Trusted in Peer-Reviewed Scientific Journals, News Media, and Funded Grant Applications

Examples:

- JAMA Oncology: <https://jamanetwork.com/journals/jamaoncology/fullarticle/2813585>
- BMC Public Health: <https://bmcpublihealth.biomedcentral.com/articles/10.1186/s12889-023-16787-1>
- TODAY: <https://www.today.com/health/news/covid-wave-2024-rcna132529>
- Forbes: <https://www.forbes.com/sites/judystone/2023/12/01/cdc-improves-their-covid-19-reporting-with-a-new-wastewater-dashboard>
- Salon: <https://www.salon.com/2023/10/19/a-lapse-in-wastewater-detection-is-worrying-scientists-about-distorted-data/>
- The New Republic: <https://newrepublic.com/article/177849/biden-democrats-covid-pandemic-2024>
- Yahoo! News: <https://news.yahoo.com/us-sees-largest-covid-wave-165811252.html>
- Washington Post: <https://www.washingtonpost.com/health/2024/01/12/covid-surge-january-2024/>
- Time: <https://time.com/6554340/covid-19-surge-2024/>
- Grants funded intramurally by Tulane University and externally by the American Cancer Society

Forecast for the Longer-Term – Annual Trends

*This section is being reshared from a prior report, as some people may need to make decisions about travel, surgery, and other important events several months out.

Is COVID-19 "seasonal"?

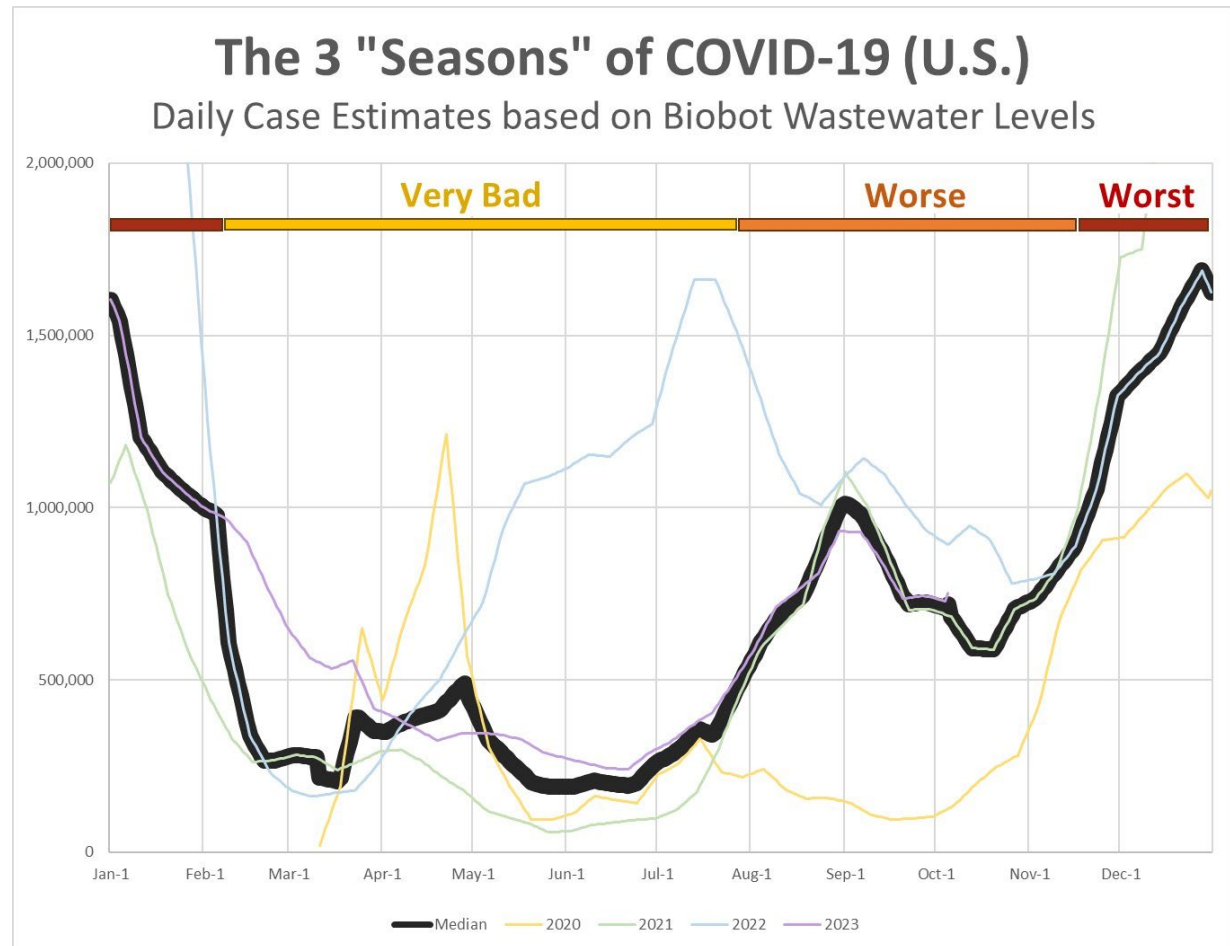
Not in any meaningful sense of the word. The following graph uses historical Biobot #wastewater data to estimate daily case rates using the PMC model. Thin lines show 2020 (yellow), 2021 (green), 2022 (blue) and 2023 (purple). The black line is the median. It is not really a forecast, but merely a summary of historical data. To the extent the median provides a reasonable approximation of the future, it is a useful starting point for a gift-level forecast.

Season 1: "Very Bad Transmission"

Focusing on the median, you'll see that case rates tend to be lower (but still in the 250-500K/day range) from mid-Feb through the end of July. These are valuable data. If I needed to schedule a non-urgent surgery, when would I do it? Late February, when transmission has often dropped, but before the general public not monitoring wastewater has realized so, perhaps meaning some people are still using airborne precautions. You're basically hopefully beating the transmission "market." I'd also be prepared to cancel an appointment or push back 6 weeks if needed.

Season 2: "Worse Transmission"

Again, focusing on the median, you see a late summer wave from August through mid-October. This is the clearest indication that C19 is not "seasonal," if people are using that term to mean an annual event. If we were doing two boosters a year, it seems like booster 1 would roll out in July. Why do we have this wave? Schools have very little mitigation (poor air quality, little/no testing, little/no masking, low vax rates). Also, the



fleeting immunity from winter boosters and infections has waned. If I had an urgent maskless medical/dental visit, I'd schedule mid-October through early November and cross my fingers (around my HEPA). It's still high transmission but about to get worse. This is also a good time to stock up on N95s, rapid tests, and HEPA filters before the prices may increase, scarcity may become a problem, or one has an infection in the home. Travel insurance is wise.

Season 3: "Worst Transmission"

From mid-November to mid-Feb, transmission is extremely problematic, according to the median line. Everybody should be wearing high-quality masks, testing as frequently as possible, improving indoor air quality, and moving activities outdoors and remote.

A Couple Caveats

Seasonality. Some people use the word "seasonal" to mean predictable, rather than merely a discrete 2-3 month season of transmission. In some ways, transmission is predictable. You'll see the 2023 purple line has followed the median very closely. However, we're talking about a very small sample size of years, so one would expect one of the years to mostly follow the median. Also, there are clear discrepant cases. BA.1 goes off the chart (winter 2021 to early 2022). The 2022 summer wave was also sizable. My approach is to make longer-term plans based on the median line and then be prepared to shift plans toward more remote activities if a large wave picks up. Hopefully, transmission becomes more predictable as years go by, but I'm not betting on it yet.

Case estimates. If you have followed the PMC dashboard, you'll know these are estimated by linking Biobot wastewater levels to IHME true case estimates. I would find case estimates 15% higher or 30% lower also reasonable and discuss these estimates with many modeling experts. There are also some more sophisticated models, where I believe an argument can be made that waves are actually marginally more leptokurtic (spikier mountains and deeper valleys than shown here).

General Technical Notes, Not Specific to the Current Week's Report

Status of Biobot wastewater reporting

The estimates and forecast described here use wastewater data reported by Biobot. Biobot is now updating their data on Fridays or Mondays, and the CDC has awarded several prior Biobot sites to a company called Verily. The transitional phase at Biobot seems mostly through, though Biobot is contesting the contract reassignment in court. As long as national wastewater data are being reported, the PMC reports will continue.

Case estimates

Case estimates were used by evaluating various potential multipliers to go from wastewater levels to cases. To identify true cases, not merely just reported cases, I used the IHME's case estimates for January 1, 2021 through April 1, 2023 (<https://covid19.healthdata.org/united-states-of-america?view=cumulative-deaths&tab=trend>). I compared wastewater with their case estimates on the 1st of each month. The correlation was $r=.94$. The maximum possible correlation is 1.00, so that is freakishly high, higher than just about any of the 10,000 or so correlations I've ever run. I was hoping for a correlation of $r=.70$ or higher, which still would have been great. Basically, wastewater is a supreme indicator of case rates. Next, I examined multipliers. Are cases 10x the arbitrary wastewater metric? 10,000x? Something else? Take cases and divide by wastewater at each data point, then find a summary metric (mean, median, trimmed mean, etc.). The metric I found most defensible was to use a +/-10% trimmed mean (average that excludes extreme data points, where case estimates are more error-prone), where each unit of wastewater translated into 1455 cases. I would find multipliers of 1000 to 1700 (31% lower to 17% higher) also reasonable. Arguably, case rates are magnitudes (10-100 times) higher than many people expect, so these details have minimal practical significance for everyday decision making. There are also more sophisticated strategies, such as regression models, but I found those results to be counter-intuitive (e.g., positive intercept, where I would have expected zero or negative). One can set the intercept to zero, use various heteroscedasticity-related techniques, and correct for the lack of imperfect reliability, but most of that is over the heads of people using this model and would accomplish little more than the trimmed multiplier method. The multiplier method has also led to techniques (only posted on Twitter thus far) for making regional estimates using very simple multipliers. Elegant is good.

Percentage infectious

After estimating the current number of new infections, it is relatively straightforward to estimate the percentage of the U.S. population actively infectious with COVID-19, but there are several caveats worth noting. One, the U.S. population is assumed to be 334,565,848. This was the CDC-estimated U.S. population on the final day of the IHME case estimation model. The number of new daily cases divided by the population tells one the percentage of the population newly infected today, often small at around 0.3% or less. Two, consider the infectious window. The percentage of the population infectious depends on the percentage of new people infected but also the duration people stay infectious. The model assumes people stay infectious for 7 days. Low estimates are that people are infectious for an average of 5 days (this defies the preponderance of the evidence, in my view), and high estimates are more like 10 days (too high in my view, based on a preference for round numbers). Other compelling estimates are more like 8-8.5 days. This duration may change over time, based on new variants, new vaccines, vaccine utilization rates, and treatments. If assuming the infectiousness duration is 10% longer, multiply by 1.10. If assuming 20% shorter, multiply by 0.80. New cases divided by the population equals new daily infections. Note also, these are merely averages and do not reflect individual variation, as some get infected and

are not contagious, whereas others get infected and remain infectious likely for months (extremely rare). New daily infections multiplied by the number of days infectious indicates the percentage of the population actively infectious.

Long COVID

Long COVID case estimation. The lower and upper bounds for Long COVID case estimates assume that 5-20% of people infected with SARS-CoV-2 will develop Long COVID as a result of that infection. Some published reports and analysts have suggested lower (1%) or higher (40%) values. A useful framework for thinking about these estimates is that the low value is more indicative of people experiencing serious, enduring, known harms, whereas the upper estimates are closer to the number experiencing disruptive symptoms for at least several months, perhaps with full or partial recovery. These estimates do not indicate unknown long-term harms. For example, if infections increase the risk of cancer or cardiovascular disease substantially and with increasing risk over 10-30 years, that is not captured well by these metrics. The metrics also do not encompass the 1.2 to 1.8 million Americans who have died of COVID-19. Future models may incorporate estimates of mortality. Finally, the estimates project the number who will ultimately experience Long COVID from a new infection, but that is several months down the line. The estimates reflect future implications. For simplicity of interpretation, they are not modeling the number of new Long COVID cases today that resulted from infections three months ago.

General forecasting model specification

The forecasting models are elegant, meaning simple and effective. In regression analyses using historical pandemic wastewater data, the model explains 96% of the variance in the following week's forecast. The model is simple. It includes the year (2020, 2021, 2022, or 2023). It includes the historical median (switched from average on 12/11/23) for the current half month; imagine the year sliced into 26 pieces, and it incorporates data on the historical median for that half month (e.g., second half of September). The model also incorporates four lagged variables, the wastewater levels 1, 2, 3, and 4 weeks ago. Overall, you can think of the model as having two main processes. One incorporates what we know historically. The other incorporates what has been happening the past several weeks. The historical data are useful because transmission mostly, but not always, follows a particular monthly pattern. It is not seasonal in that there are not just three bad months a year, but there is month-to-month variation, and sometimes even useful differences between the first versus second half of the month. The use of recent wastewater estimates helps in several ways. It lets the model know if something about the current point in time differs dramatically from the historical data, and it quickly adapts the model to changes, such as if a wave is starting or ending,

Real-time model (purple line)

This model assumes that real-time data reports of wastewater levels are accurate. However, real-time data often get corrected. Some sites may be slow reporting, and if there is a bias built in, such as places with high transmission being late to report, that would be a problem. Often, the real-time reports are quite accurate, but occasionally they have been corrected substantially a week later. The general model places a lot of weight on the most recent data, so any errors here can lead the model to assume a wave is picking up that really is not (false alarm) or that things are improving better than expected (false hope).

Alt model #1, turtle (green line)

The turtle model moves slow and steady. It completely ignores the most recent week's worth of data from Biobot, treating it as unreliable. It will ignore false fluctuations inferred from inaccurate real-time reporting. However, it will be slower to respond to real changes, such as the onset in a new wave or the decline in a wave that has peaked.

Alt model #2, cheetah (orange line)

The cheetah model moves fast. It aims to correct for biases in real-time data reports. If last week's real-time report overestimated levels by 10% upon correction, it assumes this week's real-time report suffers the same bias. If last week's real-time report underestimated true levels, it assumes the same for this week. If last week's real-time report was accurate, it will look similar to the real-time model. This model is very good if there is a bias, such as if areas with high transmission experience delays in reporting. However, it can also be overreactive. If there was some error in a real-time report that was just "random" rather than biased in a particular correction, it will tend to overcorrect the next week's model.

Composite Model (red line)

This is the arithmetic average of the three models. It's what's used for deriving all of the statistics reported. When all of the individual models are very close to the average, that suggests high confidence. When the models make vastly different predictions, that suggests more uncertainty in the data, largely based on perceptions of the accuracy of real-time wastewater reporting.