

COVID-19 Counts in Washington State and Select Other Locations

Nathan (Nat) Goodman

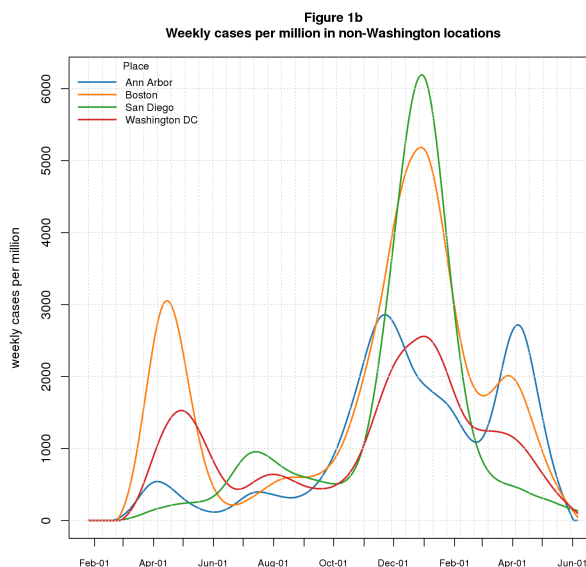
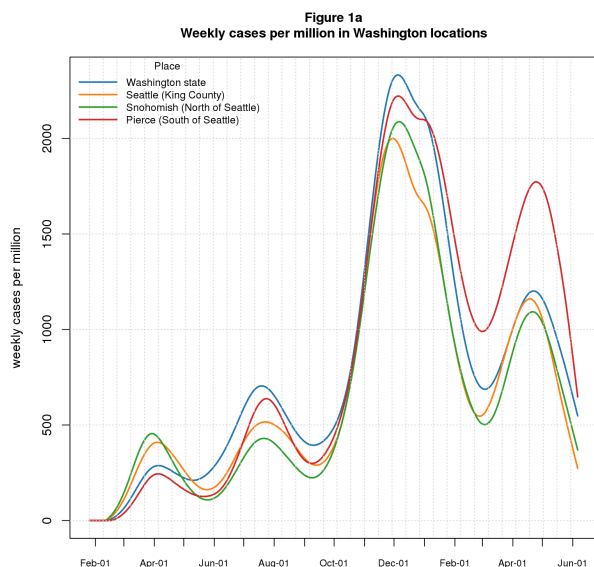
June 16, 2021

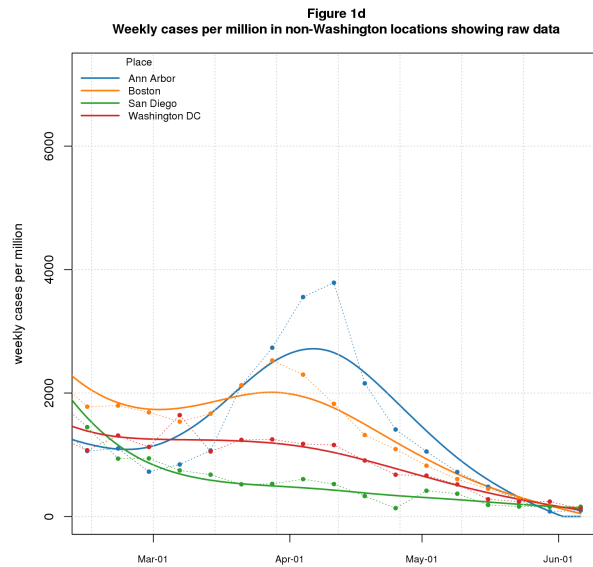
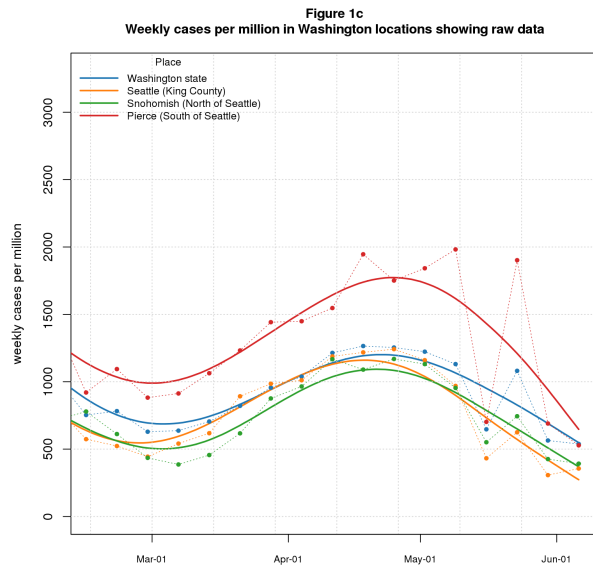
Counts of cases, hospital admissions, and deaths are key metrics of COVID-19 prevalence and burden, and are the basis for model-based estimates and predictions of these statistics. I present here graphs showing these metrics over time in Washington state and a few other USA locations of interest to me. I update the graphs and this write-up weekly. Previous versions are [here](#). See below for [caveats](#) and [technical details](#).

Figures 1a-d show smoothed *case* counts per million for several Washington and non-Washington locations. The Washington locations are the entire state, the Seattle area where I live, and the adjacent counties to the north and south (Snohomish and Pierce, resp.). The non-Washington locations are Ann Arbor, Boston, San Diego, and Washington DC.

Figures 1a-b (the top row) show smoothed data (see details [below](#)). Figures 1c-d (the bottom row) overlay raw data onto the smoothed since mid-February to illustrate recent variability.

The figures use data from Johns Hopkins Center for Systems Science and Engineering (JHU), described [below](#). When comparing the Washington and non-Washington graphs, please note the difference in y-scale: the current raw counts for Washington locations (about 350-650 per million) are much greater than for non-Washington (about 0-150 per million).

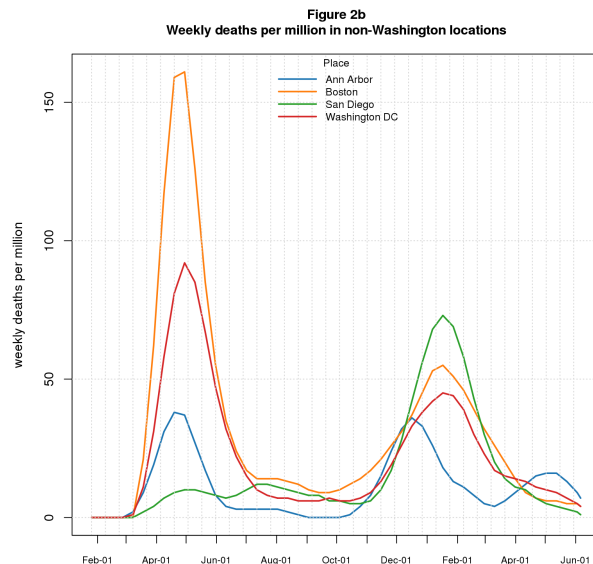
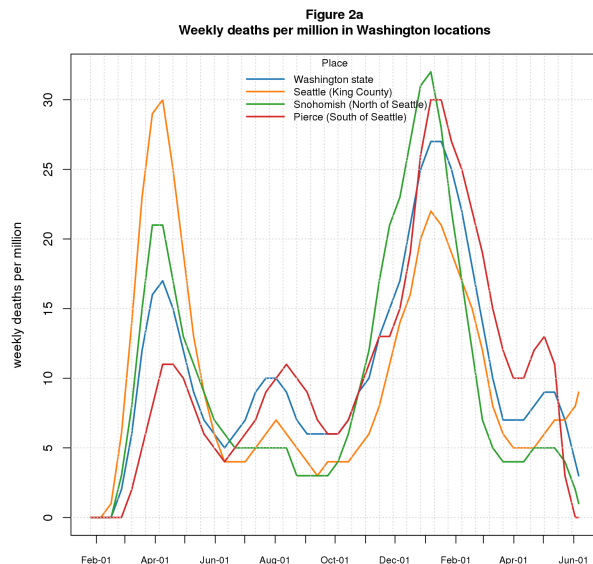


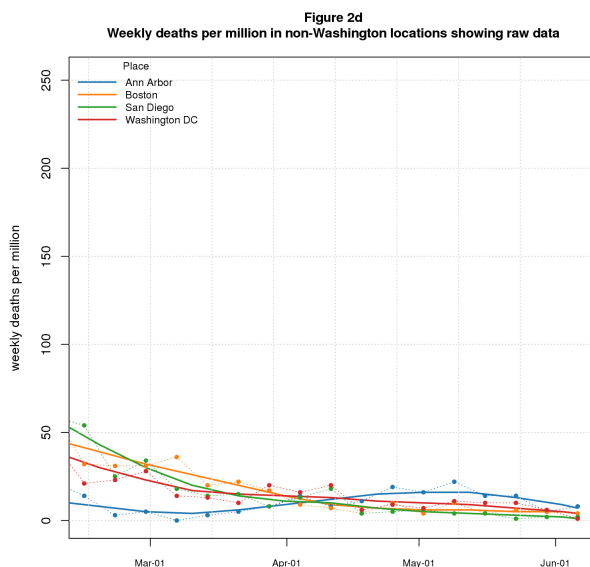
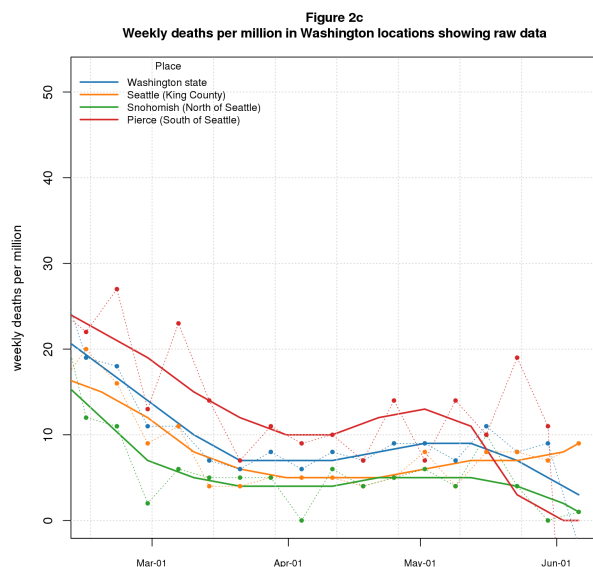


The smoothed graphs for Washington (Figure 1a) show that rates are declining in all locations. Trend analysis supports the decline over the past six and eight weeks; raw data (Figure 1c) and trend analysis suggest that rates are flattening.

The picture for non-Washington locations (Figure 1b) is even better: cases are declining everywhere. Trend analysis and raw data (Figure 1d) confirms the decline everywhere except San Diego where rates seem to be flattening at a low level.

Figures 2a-d show *deaths* per million for the same locations. When comparing the Washington and non-Washington graphs, again please note the difference in y-scale: the current Washington rates are 0-9 per million; the non-Washington rates are 1-7 per million. As with the *case* graphs, the top row (Figures 2a-b) show smoothed data (see details [below](#)) and the bottom row (Figures 2c-d) overlays raw data onto the smoothed since mid-February.



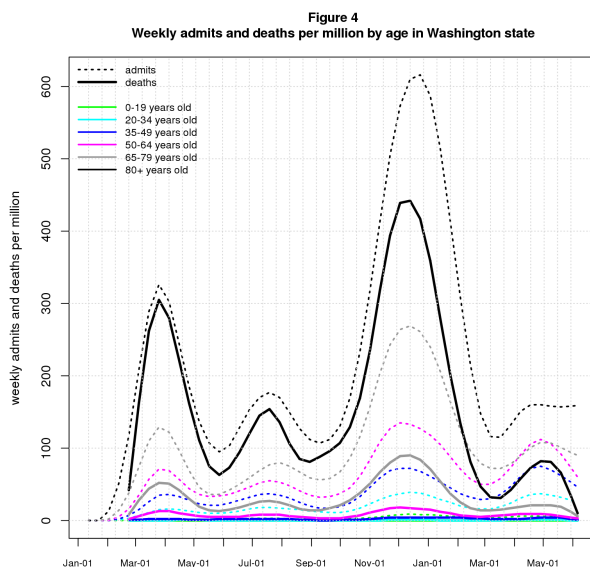
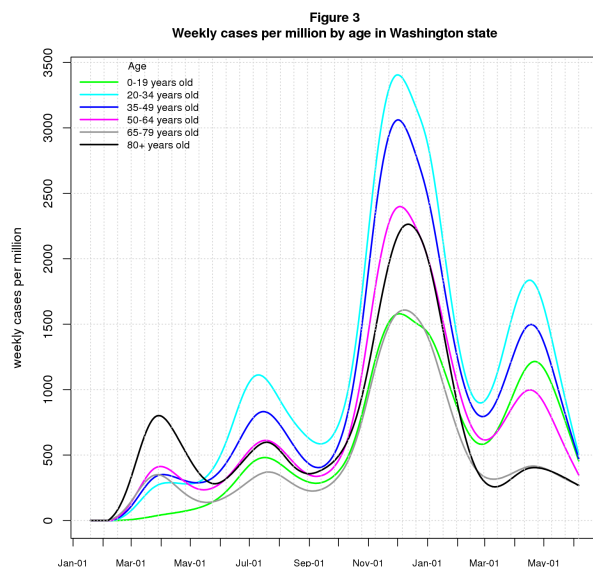


The smoothed Washington data (Figure 2a) shows three waves. The second peak was thankfully lower than the first; the third wave exceeded the first in all areas except Seattle (King County). The graphs are well down from the first and third peaks, and are finally below their summer 2020 peaks everywhere except Seattle. The graphs are heading down everywhere except Seattle; trend analysis and raw data (Figure 2c) indicate the data is too variable to be confident in the direction.

The smoothed non-Washington data (Figure 2b) shows early peaks in most locations, followed by a long trough, followed by a second wave starting in November 2020. The graphs are well down from their recent peaks. Deaths are decreasing or flat everywhere. Trend analysis and raw data (Figure 2d) suggest that rates have flattened at low levels.

The next graphs show the Washington results broken down by age. This data is from Washington State Department of Health (DOH) weekly downloads, described [below](#). An important caveat is that the DOH download systematically undercounts events in recent weeks due to manual curation. I extrapolate data for late time points as discussed [below](#).

Figure 3 is *cases*. Figure 4 shows hospital admissions (*admits*) and *deaths*.



Early on, the pandemic struck older age groups most heavily, but *cases* quickly spread into all age groups, even the young. During the second wave, young adults and middle aged adults (20-34 and 35-49 years) became the most affected groups. The third and fourth waves swept into all age groups. Thankfully, the fourth wave seems to have spared the oldest age groups (65-79 and 80+) perhaps due to the high vaccination coverage in these people.

The shocking *death* rate of the 80+ age group jumps off the page in Figure 4. The *death* rate for this age group shows four waves. The third wave reached its peak in December 2020. The fourth wave peaked in April 2021 at a level well below the earlier peaks. Recent DOH counts are well below JHU (see [Figure 5b below](#)), though my extrapolation and fitting are bringing the processed DOH data close to JHU. Throughout the pandemic, the death rate for the 80+ group was much higher than any other group, but it now appears that deaths even in this group have dropped to near zero. Given the declining *case* rates, it seems reasonable to be optimistic that deaths will remain low in the coming weeks.

The *admit* rate shows the same four waves, and seems to be decreasing in most groups. (But again, heed the [caveat](#) about DOH undercounting).

In versions before April 28, 2021, Figure 4 only showed *deaths*, but the growing impact on young people combined with the high vaccination rate of the old, make admissions a useful indicator of pandemic status. These earlier versions aggregated 0-64 years into a single group, since the *death* rate in these ages is near 0. Now that the graph includes *admits*, it's useful to keep the age groups separate.

Caveats

1. The term *case* means a person with a detected COVID infection. In some data sources, this includes “confirmed cases”, meaning people with positive molecular COVID tests, as well as “probable cases”. I believe JHU only includes “confirmed cases” based on the name of the file I download.
2. Detected *cases* undercount actual cases by an unknown amount. When testing volume is higher, it's reasonable to expect the detected count to get closer to the actual count. Modelers attempt to correct for this. I don't include any such corrections here.
3. The same issues apply to *deaths* to a lesser extent, except perhaps early in the pandemic.
4. The geographic granularity in the underlying data is *state* or *county*. I refer to locations by city names reasoning that readers are more likely to know “Seattle” or “Ann Arbor” than “King” or “Washtenaw”.

5. The date granularity in the graphs is weekly. The underlying JHU data is daily; I sum the data by week before graphing.
6. I truncate the data to the last full week prior to the week reported here.

Technical Details

1. I smooth the graphs using a smoothing spline (R's `smooth.spline`) for visual appeal. This is especially important for the *deaths* graphs where the counts are so low that unsmoothed week-to-week variation makes the graphs hard to read.
2. The Washington DOH data (used in Figures 3 and 4 to show counts broken down by age) systematically undercounts events in recent weeks due to manual curation. I attempt to correct this undercount through a linear extrapolation function (using R's `lm`). I have tweaked the extrapolation repeatedly, even turning it off for a few weeks. The current version uses a model that combines date and recentness effects. In past, I created models for each Washington location and age group but had to change when DOH changed its age groups on March 14, 2021 (see [below](#)). I now create a single model for the state as a whole and all age groups summed together, then blithely apply that model to all locations and ages.
3. The trend analysis computes a linear regression (using R's `lm`) over the most recent four, six, or eight weeks of data and reports the computed slope and the p-value for the slope. In essence, this compares the trend to the null hypothesis that the true counts are constant and the observed points are randomly selected from a normal distribution. After looking at trend results across the entire time series, I determined that p-values below 0.1 indicate convincing trends; this cutoff is arbitrary, of course.
4. In previous versions of the document, I've shown plots that overlay raw *case* and *death* counts onto the smoothed graphs in Figures 1 and 2. Recent data is much less variable than in past making these plots superfluous.
5. In most versions before March 17, 2021, I showed counts broken down by age (as in Figures 3 and 4) graphs for each Washington location. Now I only show the statewide graphs: the other locations are similar.
6. In past, DOH reported results in 20 year age groups starting with 0-19, with a final group for 80+. As of the March 14, 2021 data release (corresponding to document version March 17), DOH changed age groups. The new groups are 0-19, followed by 15 year ranges (20-34, 35-49, 50-64, 65-79) with a final group for 80+.

Data Sources

Washington State Department of Health (DOH)

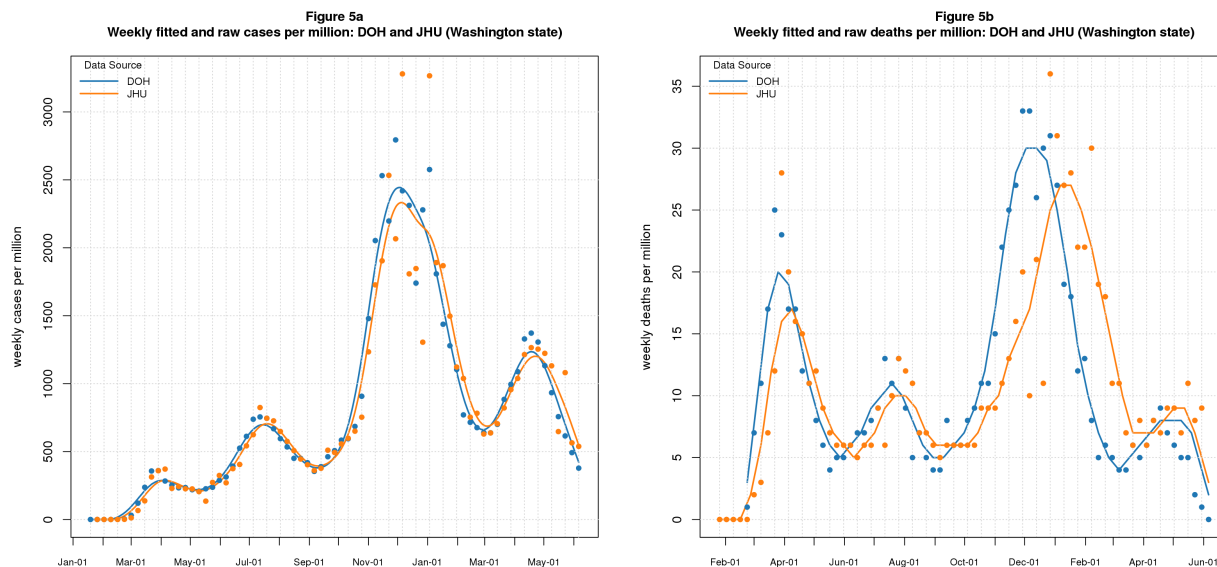
DOH provides three COVID data streams.

1. [Washington Disease Reporting System \(WDRS\)](#) provides daily “hot off the presses” results for use by public health officials, health care providers, and qualified researchers. It is not available to the general public, including yours truly.
2. [COVID-19 Data Dashboard](#) provides a web graphical user interface to summary data from WDRS for the general public. (At least, I think the data is from WDRS - they don't actually say).
3. Weekly data downloads (available from the [Data Dashboard](#) web page) of data curated by DOH staff. The curation corrects errors in the daily feed, such as, duplicate reports, multiple test results for the same incident (e.g., initial and confirmation tests for the same individual), incorrect reporting dates, incorrect county assignments (e.g., when an individual crosses county lines to get tested).

The weekly DOH download reports data by age group. In past, the groups were 20-year ranges starting with 0-19, with a final group for 80+. As of March 14, 2021, they changed the groups to an initial 20-year range (0-19), then several 15-year ranges (20-34, 35-49, 50-64, 65-79), with a final group for 80+.

Issues with DOH Extrapolation

Figures 5a-b compare DOH and JHU *cases* and *deaths* for Washington state to illustrate the undercount in the raw DOH data. The *cases* data matches well except for a few periods spanning several weeks, including most of May. The *deaths* data matches less well and is presently much lower than JHU. I have no explanation for the discrepancies.



Johns Hopkins Center for Systems Science and Engineering (JHU)

JHU CSSE has created an impressive portal for COVID data and analysis. They provide their data to the public through a [GitHub repository](#). The data I use is from the [csse_covid_19_data/csse_covid_19_time_series](#) directory: [time_series_covid19_confirmed_US.csv](#) for *cases* and [time_series_covid19_deaths_US.csv](#) for *deaths*.

JHU updates the data daily. I download the data the same day as the DOH data (now Tuesdays) for operational convenience.

Other Data Sources

The population data used for the per capita calculations is from [Census Reporter](#). The file connecting Census Reporter *geoids* to counties is the [Census Bureau Gazetteer](#).

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