COVID-19 Models: Forecasting the Pandemic’s Spread
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With our evolving understanding of the details of COVID-19 infection and transmission, the impact of social-distancing measures and other factors, hospitals and health systems face complex forecasting challenges.

A growing number of models have been developed by health care systems, academic institutions, consulting firms and others to help forecast COVID-19 cases and deaths, medical supply needs, including ventilators, hospital beds and intensive care unit (ICU) beds, timing of patient surges and more.

We recognize that all forecasting is imperfect and that models often change as data are revised to reflect evolving conditions in the population. The American Hospital Association (AHA) offers this resource in response to members’ interest in tools and resources that can help them make decisions that are consistent with their organizations’ values and community needs.

The **CASE PROJECTION AND CAPACITY PLANNING** models are all predicated on some assumptions about the current status of the disease, how quickly it spreads (e.g., doubles), and/or the degree of containment efforts being deployed. Those assumptions drive the rise, peak and decline in “the curve” — the number of COVID-19 cases, deaths or resources needed. The simpler models apply the user’s inputs on the local population and current status of COVID-19. The more sophisticated models allow the user to alter other parameters (e.g., changes in social distancing) that influence the curve. Some of the models match projected cases to existing capacity to estimate when and where a caseload surge will exceed capacity.

The **COMMUNITY VULNERABILITY MAPS** use social and economic population variables, such as density, travel distances to health care, underlying chronic illnesses, economic status, to highlight communities that are most at risk for contracting and/or succumbing to the disease.

The **CASE MAPPING AND PROJECTION TOOLS** provide a quick visual depiction of disease incidence at the national, state or county levels.

The following list of resources was updated on April 6 and is not exhaustive. It will continue to be updated periodically as additional models are assessed. These resources are classified into categories based on the extent of their modeling capabilities and how they can be used.

Please note that the tools listed do not necessarily reflect the views of the AHA, nor do they constitute advice from the AHA. Rather, they are meant to assist hospitals and health systems in making decisions for their own organizations.
CASE PROJECTION AND CAPACITY PLANNING

Penn Medicine — CHIME: The COVID-19 Hospital Impact Model for Epidemics

This model allows hospitals to enter information about their population and modify assumptions about the spread and behavior of COVID-19. It then runs a standard Susceptible-Infected-Recovered (SIR) model to project the number of new COVID-19 hospital admissions each day, along with the hospital census of COVID-19 patients. These projections then can be used to create best- and worst-case scenarios to assist with capacity planning. Penn Medicine is open-sourcing CHIME and making it available to the health care community.

- Data and assumption sources: The discrete-time SIR model has user inputs for population, the hospital’s market share, current COVID-19 hospitalizations, doubling time, severity and social distancing. It employs a combination of estimates from other locations (e.g., Wuhan, China; South Korea; Italy), informed estimates based on logical reasoning.
- Estimation level: Individual hospital or health care system.
- Degree of complexity: User interface with tutorial and documentation.
- Software platform/language: Web-based.
- Ability to change assumptions and run scenarios: Yes.
- Output: Projected new COVID-19 admissions per day (e.g., hospitalized, ICU, ventilator) and projected total COVID-19 patients per day for the same three cohorts. Projections for SIR patients.
- Methodology: SIR model documentation.

Health Catalyst — Capacity Planning Tool

This tool builds on Penn Medicine’s model. Health Catalyst added new features such as the ability to run multiple scenarios, store the scenarios on users’ local desktops, and then upload them again for later use. Additional features include predicted impacts of scenarios on hospital capacity (e.g., beds and ventilators).

- Data and assumption sources: Penn Medicine’s discrete-time SIR model. User inputs for population, market share, current COVID-19 hospitalizations, hospital capacity, doubling time, severity and social distancing.
- Estimation level: Individual hospital or health care system.
- Degree of complexity: User interface with tutorial and documentation.
- Software platform/language: Web-based.
- Ability to change assumptions and run scenarios: Yes.
- Output: Projected new COVID-19 admissions per day (e.g., hospitalized, ICU, ventilated) and projected total COVID-19 patients per day for the same three cohorts. Projections for SIR and bed capacity (e.g., total beds, ICU beds, ventilators).
- Methodology: Statistical model documentation and application guidance.
Cleveland Clinic — *Hospital Impact Modeling*

Also building on the Penn Medicine model, Cleveland Clinic’s model is available in GitHub, with documentation. It uses Penn’s SIR assumptions, but allows more flexibility to hold time periods constant and model various scenarios as well as update the curves, based on inputs such as changes in response measures. The model is Statistical Analysis System (SAS)-based and, in its current iteration, does not have a user interface. Users will need advanced SAS skills to alter assumptions.

- **Data and assumption sources:** This model extrapolates to several model types like the Susceptible-Exposed-Infected-Removed (SEIR) model. User inputs for population, market share, admission rate for COVID-19 on Day 0, COVID-19 infections on Day 0, social-distancing baseline and change, incubation period, recovered patients, recovery days, doubling time, ICU percentage, ventilator percentage, fatality rate and length of stay. Assumptions are based on the population studied in the Cleveland Clinic and SAS collaboration.
- **Estimation level:** Individual hospital or health care system.
- **Degree of complexity:** Complex SAS code with documentation. Geared to users with SAS experience.
- **Software platform/language:** SAS.
- **Ability to change assumptions and run scenarios:** Yes.
- **Output:** Daily projections of disease progression and hospitalization.
- **Methodology:** Statistical models documentation.

UW-IHME — *COVID-19 Forecasting Tool*

The University of Washington’s Institute for Health Metrics and Evaluation forecasting tool provides state-by-state information on anticipated number of beds needed, ICU beds needed and deaths related to COVID-19. Projections assume full social distancing through May. The tool does not permit hospital-level inputs and alternate scenarios.

- **Data and assumption sources:** Data on confirmed COVID-19 deaths by day from World Health Organization websites and local and national governments; data on hospital capacity and utilization for U.S. states from the AHA; and observed COVID-19 utilization data from select locations (e.g., Italy, China, Korea, and the U.S). Model forecasts peak death and resource usage dates, and daily deaths and hospital utilization (e.g., beds, ICU beds, ventilators) against capacity by state and for the U.S. over the next four months (until July), assuming full social distancing through May.
- **Estimation level:** Individual states and U.S.
- **Degree of complexity:** No user inputs, easy to comprehend.
- **Software platform/language:** Web-based.
- **Ability to change assumptions and run scenarios:** No.
- **Output:** Graphs showing peak death and resource utilization dates, daily total bed and ventilator needs and shortages, and projected deaths.
- **Methodology:** Statistical models (e.g., nonlinear mixed-effects model) documentation.
**Stanford Medicine — Hospital Bed Projections**

This COVID-19 ICU and floor-projection planning tool uses a facility’s own patient census numbers and hospital capacity. The model is designed to facilitate hospital planning with estimates of the daily number of intensive care beds, acute care beds and ventilators necessary to accommodate patients who require hospitalization for COVID-19 and how these compare to the available resources.

- **Data and assumption sources:** COVID-19 admissions are projected with the exponential growth model based on the inputs of the doubling time and the initial number of patients. The user inputs the census data of the ICU, floor and general medicine, along with daily admissions and hospital capacity parameters.
- **Estimation level:** Individual hospital or health care system.
- **Degree of complexity:** User interface has the option to upload facility’s daily census data or use built-in parameters.
- **Software platform/language:** Web-based.
- **Ability to change assumptions and run scenarios:** Yes.
- **Output:** Projected number of patients in ICU, non-ICU and on ventilators and days until capacity is exceeded.
- **Methodology:** [Model explanation in more detail](#).

**RAND Corporation — Critical Care Surge Response Strategies**

In this report, the authors present a list of strategies for creating critical care surge capacity and estimating the number of patients accommodated, given the number of available critical care doctors and nurses, respiratory therapists, ventilators and hospital beds. They also document the development of a user-friendly, Microsoft Excel-based tool that allows decision-makers at all levels — hospitals, health care systems, states, regions — to estimate current critical care capacity and rapidly explore strategies for increasing it.

- **Data and assumption sources:** Formulas driven by user inputs — staff ratios, staff levels, additional personnel resources and patient capacity; model uses AHA data, American Medical Association data, and other data on critical care space, staff and supplies (e.g., ventilators, PPEs).
- **Estimation level:** Individual hospital, health care system, regional or state levels.
- **Degree of complexity:** Clear descriptions of required inputs.
- **Software platform/language:** Excel.
- **Ability to change assumptions and run scenarios:** Yes.
- **Output:** Total number of patients who can be cared for per shift by staff and resource type.
- **Methodology:** Limited [documentation](#).
COMMUNITY VULNERABILITY MAPS

Jvion — COVID-19 Community Vulnerability Map
Clinical artificial intelligence company Jvion launched this public-facing tool, built on Microsoft Azure maps. It enables health care providers and communities battling the COVID-19 pandemic to identify the social determinants of health (SDOH) that put populations at greater risk, informing community planning and resource allocation to proactively mitigate the risk to vulnerable populations. The interactive map identifies populations down to the census block level that are at risk for severe outcomes upon contracting a virus like COVID-19.

Conduent — Locating At-Risk Populations Tool
Conduent Healthy Communities Institute specializes in providing SDOH data to organizations so that they can improve health in their respective populations. Their tool lets users track at-risk populations by state and county.

CASE MAPPING AND PROJECTION TOOLS

Oliver Wyman Consulting — COVID-19 Scenario Generator
This scenario generator from Oliver Wyman predicts the growth and peak of COVID-19 cases in the U.S. at the county level, based on the effectiveness of and time to implement containment measures. The tool allows for some user input and scenario modeling.

Dalla Lana School of Public Health, University of Toronto — COVID-19 Growth Model
The goal of this model is to provide insight into the changing nature of case findings and epidemic growth.

COVID Act Now — Hospital Overload Projections for State/County Level
The U.S. Intervention Model is a data platform that projects COVID-19 infections, hospitalizations and deaths across the U.S., as well as how public health interventions contain the spread of COVID-19. It helps decision-makers understand when and how COVID-19 will affect their communities so that they can make better decisions that save lives.