

# **Australian Health Protection Principal Committee**

The **Australian Health Protection Principal Committee** is the key decision-making committee for national health emergencies. It comprises all state and territory Chief Health Officers and is chaired by the Australian Chief Medical Officer.

#### **COVID-19** forecasting group

- Peter Dawson
- Nick Golding
- Rob J Hyndman
- Dennis Liu
- James M McCaw

- Jodie McVernon
  - Pablo Montero-Manso
- Robert Moss
- Mitchell O'Hara-Wild
- David J Price

- Joshua V Ross
- Gerry Ryan
- Freya M Shearer
- **■** Tobin South
- Ruarai Tobin

#### **Data sources**

- Case-level data of all positive COVID-19 tests: onset and detection times.
- Daily population mobility data from Google, Apple & Facebook
- Weekly non-household contact surveys
- Weekly behavioural surveys
- Daily case numbers from many countries and regions via the Johns Hopkins COVID-19 repository

#### **Case numbers**

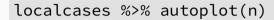
##

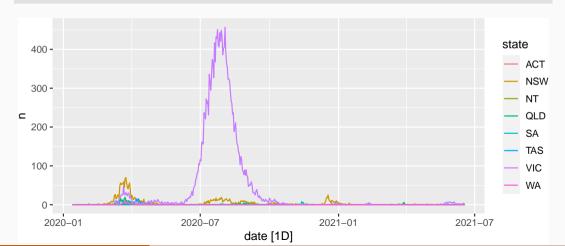
9 2020-07-09 VTC

```
localcases %>% filter(state == "VIC", date >= "2020-07-01")
## # A tsibble: 353 x 3 [1D]
## # Kev: state [1]
     date state
##
                         n
##
  <date> <chr> <dbl>
##
   1 2020-07-01 VIC
                   116
##
   2 2020-07-02 VIC
                       113
##
   3 2020-07-03 VIC
                   161
##
   4 2020-07-04 VIC
                       161
##
   5 2020-07-05 VIC
                       156
                       237
##
   6 2020-07-06 VIC
##
   7 2020-07-07 VIC
                       235
##
   8 2020-07-08 VIC
                       221
```

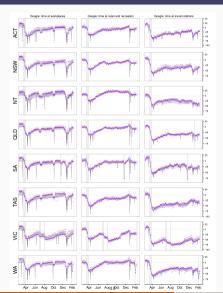
272

#### **Case numbers**





# Google mobility data



Percentage change compared to pre-COVID-19 baseline for:

- (a) time at workplace;
- (b) time at retail/recreation;
- (c) time at transit stations.

Vertical lines: physical distancing measures implemented.

# Facebook mobility data **Proportion of Facebook users** who "stayed put", 29 Feb -2 Jul 2020. Each line is one LGA. 0.4 proportion staying put 0.1

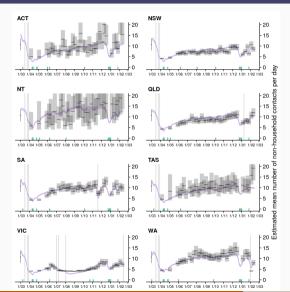
May

date

Jun

Mar

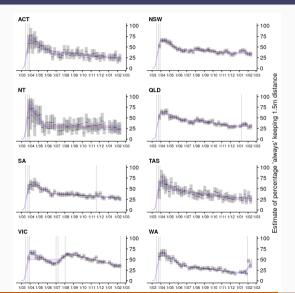
#### Macrodistancing



**Estimated # non-household contacts per day** based on nationwide weekly surveys (gray) and Google mobility data.

Green: public holidays.

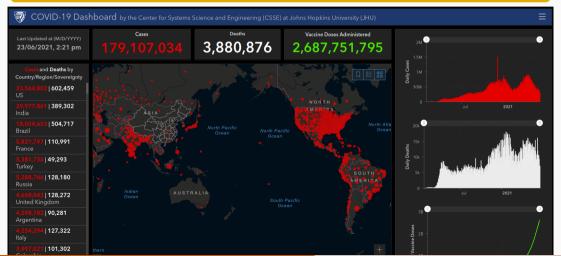
#### Microdistancing



Estimated % keeping 1.5m distance from non-household contacts based on nationwide weekly surveys (gray).

## Global daily cases by region from Johns Hopkins

#### https://github.com/CSSEGISandData/COVID-19



## Model 1: SEEIIR (Uni Melbourne/Doherty Institute)

- Stochastic susceptible-exposed-infectious-recovered compartmental model that incorporates changes in local transmission potential via a time-varying effective reproduction number.
- Uses mobility and survey data and case onset and detection times.
- $lue{}$  Daily counts  $\sim$  Negative Binomial.
- lacksquare Time in class  $\sim$  Gamma.
- Forecasts obtained using a bootstrap particle filter.

#### Model 2: Generative model (Uni Adelaide)

- Uses mobility and survey data and case onset and detection times.
- Three types of infectious individuals: imported, asymptomatic, symptomatic
- $lue{}$  Class counts  $\sim$  Negative Binomial.
- Incubation times  $\sim$  Gamma.
- Estimation via Hamilton Monte Carlo
- Forecasts obtained via simulation

# Model 3: Global AR model (Monash)

- Uses Johns Hopkins data from countries and regions with sufficient data.
- Series with obvious anomalies (negative cases and large step changes) removed.
- $n_{t,i}$  = daily cases on day t in country/region i (scaled so all data have same mean and variance).
- $y_{t,i} = \phi_1 y_{t-1,i} + \cdots + \phi_p y_{t-p,i} + \varepsilon_{t,i}$ where  $y_{t,i} = \log(n_{t,i} + 0.5)$  and  $\varepsilon_{t,i} \sim N(0, \sigma_i^2)$ .
- No stationarity constraints. Common coefficients.
- Current model has p = 24 (selected to minimize the 7-day-ahead MAE on recent Australian data).

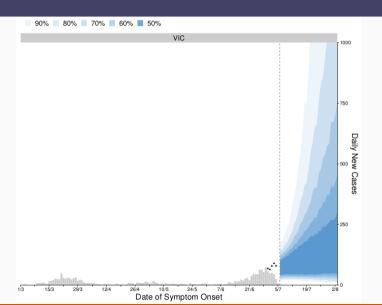
# Forecasting ensemble

 Forecasts obtained from a mixture distribution of the component models.

$$\tilde{p}(y_{t+h}|I_t) = \sum_{k=1}^{3} w_{t+h|t,k} p_k(y_{t+h}|I_t)$$

where  $p_k(y_{t+h}|I_t)$  is the forecast distribution from model k,  $I_t$  denotes the data available at time t and the weights  $w_{t+h|t,k} > 0$  sum to one.

- Also known as "linear pooling"
- Works best when individual models are over-confident and use different data sources.



 $f_{p,t}$  = quantile forecast with prob. p at time t.  $y_t$  = observation at time t

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#### Quantile score

$$Q_{p,t} = \begin{cases} 2(1-p)|y_t - f_{p,t}|, & \text{if } y_t < f_{p,t} \\ 2p|y_t - f_{p,t}|, & \text{if } y_t \ge f_{p,t} \end{cases}$$

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- Low  $Q_{p,t}$  is good
- Multiplier of 2 often omitted, but useful for interpretation
- $Q_{p,t}$  like absolute error (weighted to account for likely exceedance)
- Average  $Q_{p,t}$  over p = CRPS (Continuous Rank Probability Score)

#### **More information**

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