

Time series estimation and forecasting of Covid in Norway

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Preliminary.

Introduction

- Covid time series are typical examples of changing data-generating processes,
 - ▶ because of mutations and policy responses
 - ▶ over time and across economies
- Likely that constant-parameter models will fail.
- An approach with exogenous shocks/breaks and effects of corresponding policy responses might be a useful complementary addition to the toolbox.

The model

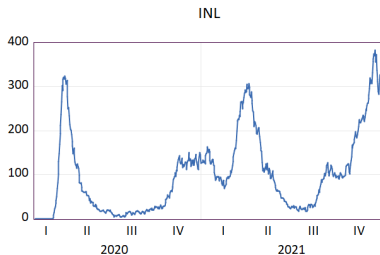
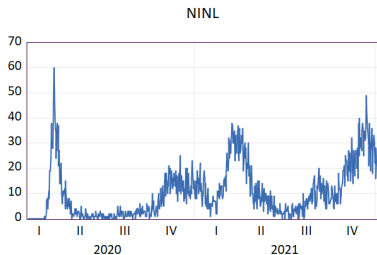
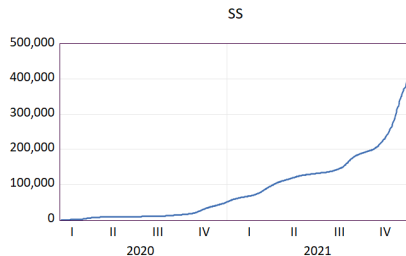
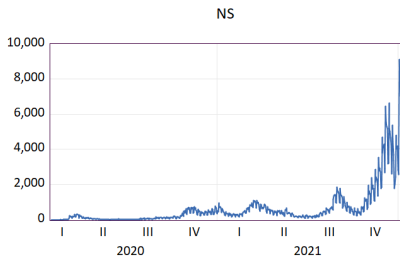
- Builds on Nymoen (2022).
- The model contains four endogenous variables:
 - ▶ NS_t , number of new infected with Covid-19, day t .
 - ▶ SS_t , accumulated number of new infected with Covid-19, day t .
 - ▶ $NINL_t$, number of new hospitalisations with Covid-19, day t .¹
 - ▶ INL_t , number of hospitalisations with Covid-19, day t .²
- The model belongs within the class of autoregressive systems. It has a recursive solution: First NS_t and SS_t are decided. Conditional upon these outcomes, $NINL_t$ is determined and finally INL_t is solved for.

¹<https://www.fhi.no/sv/smittsomme-sykdommer/corona/dags--og-ukerapporter>

²<https://www.helsedirektoratet.no/statistikk/antall-innlagte-med-pavist-covid-19-for-nedlasting>

²<https://www.helsedirektoratet.no/statistikk/antall-innlagte-med-pavist-covid-19-for-nedlasting>

The data



Use two standard models of breaks

Dummy model and smooth transition model

- Model:

$$y_t = \beta_0 + \beta_1 x_t + \beta_2 x_t D_t + \beta_3 x_t G_t + u_t$$

- Exogenous event makes binary variable D_t change effect of x_t on y_t .
- Smooth transition model³ part $G(s_{t-k}; \gamma, r)$ that changes smoothly from 0 to 1 with increasing policy target variable s_{t-k} .
- Hospitalisations INL used as policy target variable.
- The transition function $G(INL_{t-k}; \gamma, r)$ is the logistic specification

$$G(INL_{t-k}; \gamma, r) = \frac{1}{1 + \exp[-\gamma(INL_{t-k} - r)]}$$

- the threshold r is estimated.
- The parameter γ decides the steepness of the transition function.

Use two standard models of breaks (cont.)

Dummy model and smooth transition model

- If $INL_{t-k} = r$, then

$$G(INL_{t-k}; \gamma, r) = 0.5.$$

- with $(INL_{t-k} - r) \rightarrow -\infty$

$$G(INL_{t-k}; \gamma, r) \rightarrow 0$$

- with $(INL_{t-k} - r) \rightarrow \infty$

$$G(INL_{t-k}; \gamma, r) \rightarrow 1.$$

³See f. ex. van Dijk, Teräsvirta, and Franses (2002).

Breaks and regime shifts with policy response

- Model abrupt exogenous events with D .
- Model *effects* of policy response with G .
- Covid:
 - ▶ Abrupt exogenous events, like mutations (D).
 - ▶ *Effects* of policy response as target variable increases (G).
 - ▶ Development of pandemic modeled and forecasted as
 - 1 Breaks in dynamic time series model.
 - 2 Use INL_{t-5} as target variable.
 - 3 Effects of policy response when hospitalisations approach threshold level r .

The NS equation

The equation for NS_t is estimated as:

$$\begin{aligned} NS_t = & \underset{(0.003)}{0.060} (SS_{t-1} - SS_{t-14}) \\ & + \underset{(0.0009)}{0.007} (SS_{t-1} - SS_{t-14}) D_t \\ & - \underset{(0.003)}{0.007} (SS_{t-1} - SS_{t-14}) G_t \\ & + \text{lagged}(\Delta NS_{t-j}) + \text{residual} \end{aligned} \tag{1}$$

$T = 15.2.2020 - 5.1.2022, 691 \text{ obs.}$

where

$$D_t = f(\textit{Tyrol}, \textit{Alpha}, \textit{Delta}, \textit{Omicron})$$

$$G_t = \frac{1}{1 + \exp \left[\underset{(0.028)}{-0.038} \left(INL_{t-5} - \underset{(22.58)}{294} \right) \right]}.$$

The NS equation (cont.)

By definition the equation for SS_t is:

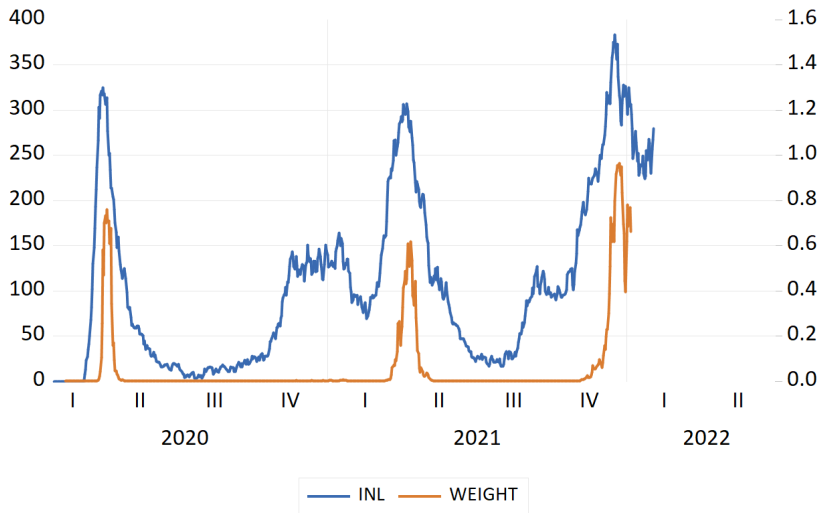
$$SS_t = NS_t + SS_{t-1},$$

The scale of incidence will be a function of the infection level in the population, which is unobservable. In equation (1) the change in the accumulated level of cases over a two-week period, $(SS_{t-1} - SS_{t-14})$ is used as an indicator of the infection level. Note that

$$(SS_{t-1} - SS_{t-14}) = \sum_{j=1}^{14} NS_{t-j}, \quad (2)$$

so (1) is an autoregressive model.

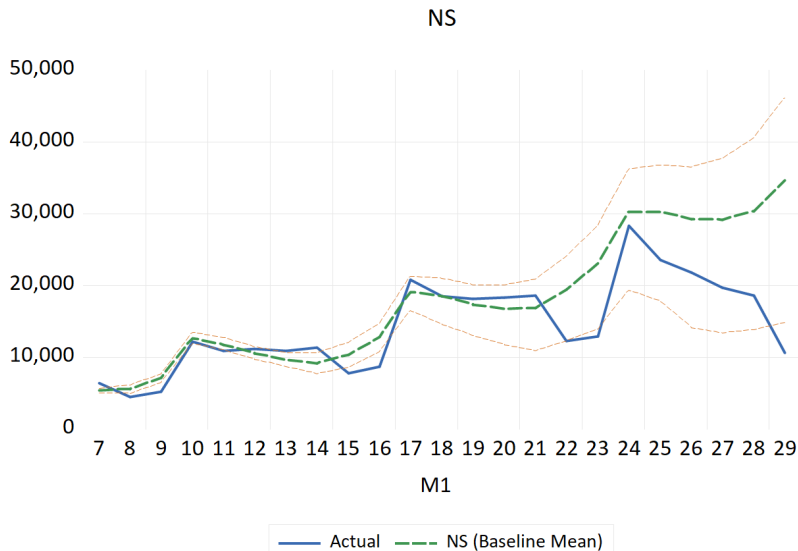
Hospitalisations INL_t and smooth transition function



tres

The forecasts and realizations of *NS*

Forecast intervals including parameter uncertainty



The *NINL* and *INL* equations

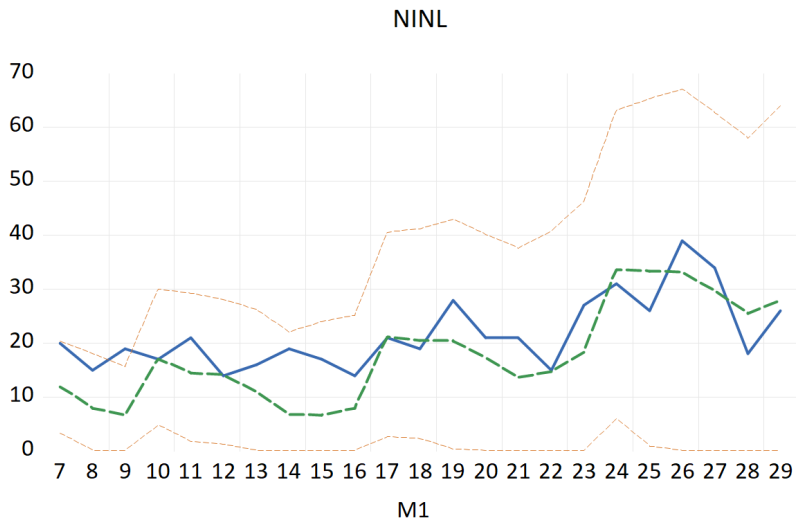
$$\begin{aligned} NINL_t = & \underbrace{-0.0006}_{(0.0001)} (SS_{t-3} - SS_{t-9}) Omicron_{t-6} \\ & + \underbrace{0.012}_{(0.001)} NS_t \\ & - \underbrace{0.007}_{(0.0009)} NS_t \times NINLDN_t \\ & - \underbrace{0.002}_{(0.0004)} NS_t \times NINLDN_{nov21_t} \\ & + \text{lagged}(NINL_{t-j}) + \text{residual} \end{aligned} \quad (3)$$

$$INL_t = \underbrace{0.901}_{(0.002)} INL_{t-1} + INL_t \quad (4)$$

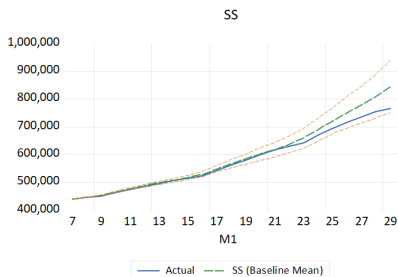
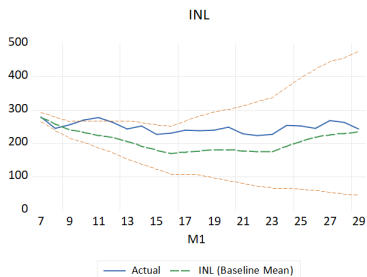
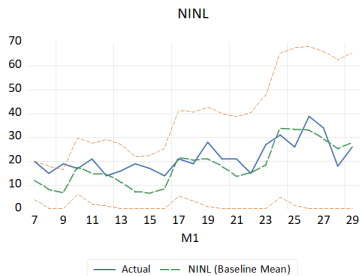
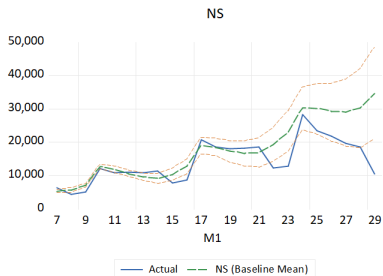
$T = 14.7.2020 - 5.1.2022$, 541 obs.

The forecasts and realizations of *NINL*

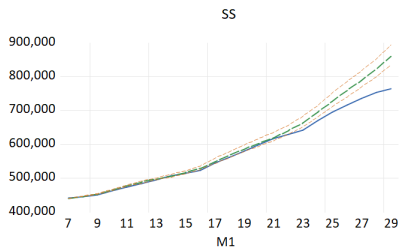
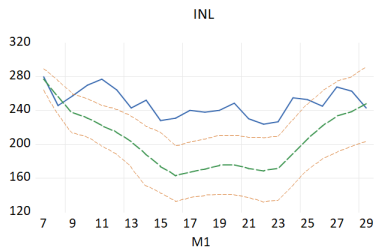
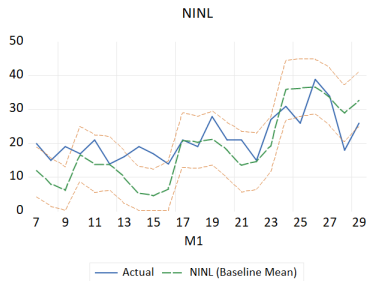
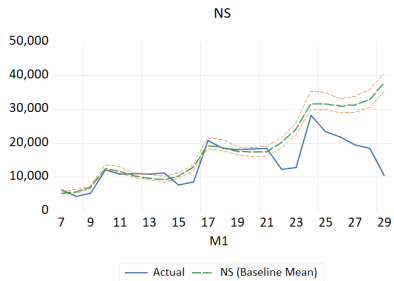
Forecast intervals including parameter uncertainty



The forecasts and realizations with parameter uncertainty

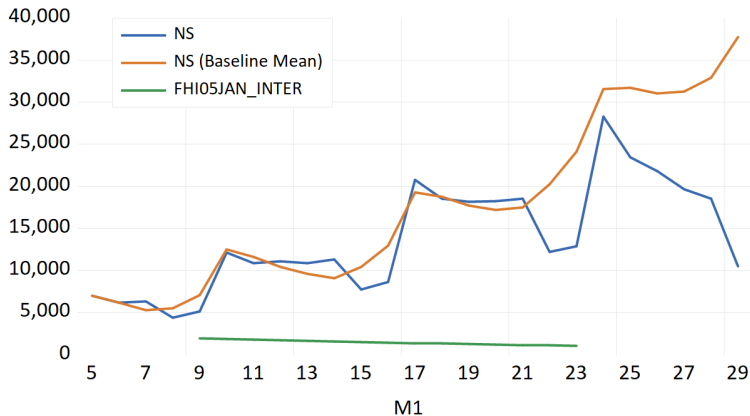


The forecasts and realizations without parameter uncertainty



Comparisons with *NIPH (FHI)*

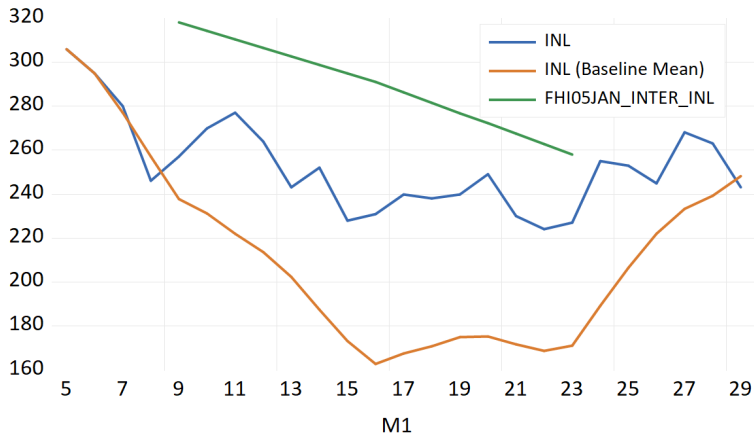
Incidents⁴



⁴ “Situational awareness and forecasting for Norway”. FHI COVID-19 modelling team Week 1, 5 January 2022. Table 2. Linear interpolation between 7, 14 and 21 days ahead forecasts from forecast origin date.

Comparisons with *NIPH (FHI)*

Hospitalisations⁵



⁵ "Situational awareness and forecasting for Norway". FHI COVID-19 modelling team Week 1, 5 January 2022. Table 2 and Figure 5. Linear interpolation between 7, 14 and 21 days ahead forecasts from forecast origin date.

Conclusions

- Covid time series are typical examples of changing data-generating processes,
 - ▶ both because of mutations and policy responses
 - ▶ both over time and across economies
- Likely that constant-parameter models will fail.
- Time series models with exogenous shifts/breaks and corresponding effects of policy response might be a useful complementary addition to the toolbox.

References

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