## Back-calculation of ICU admissions from time-series of occupied beds

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## Methods

We are interested in estimating the daily number of new ICU admissions  $I_t$  in a specific region based on the observed number of occupied ICU beds per day  $B_t = b_t$ , where t = 1, ..., T.

Let N be the number of individuals i = 1, ..., N that occupied a bed at some time-point t < T, and  $LOS_i$  be the length of the ICU stay (number of days an individual occupied a bed in the ICU) of individual I.

We assume, that the LOS follows a discrete distribution  $F_{\theta}$ , where  $P(LOS_i = k) = \theta_k$ , k = 0, ..., K and  $\sum_{k=0}^K \theta_k = 1$ . Additionally we define the probability  $P(LOS_i > k) = 1 - \sum_{l=1}^k \theta_l = \eta_k$ .

Let  $A_i$  be the admission day of individual i, and  $S_i$  be an individual-specific binary vector of length T with entry  $S_{i,t}=1$  if individual i is occupying a bed at the ICU on day t and  $S_{i,t}=0$ , if not:

$$S_{i,t} = \begin{cases} 1, & \text{if } A_i \le t \le A_i + LOS_i \\ 0, & \text{otherwise} \end{cases}$$

Based on these definitions, the observed number of occupied ICU beds on day t corresponds to

$$b_t = \sum_{i=1}^{N} S_{i,t} = \sum_{i=1}^{N} \mathbb{1}(A_i \le t \le A_i + LOS_i) = \sum_{i=1}^{N} \mathbb{1}(A_i \le t) \mathbb{1}(LOS_i \ge t - A_i)$$

and the unobserved number of new ICU admissions  $I_t$  are

$$I_t = \sum_{i=1}^{N} \mathbb{1}(A_i = t).$$

The estimation of  $I_t$  based on the observed  $B_t = b_t$  is closely related to the estimation of an inverse convolution or backprojection methods. We assume, that

the latent number of new ICU admissions on day t,  $I_t$ , are Poisson distributed with expectation  $\lambda_t$ . The observed number of occupied beds on day t are then Poisson distributed as well:

$$B_t \sim \text{Pois}(\sum_{k=0}^K \lambda_{t-k} \eta_k),$$

where  $\eta_k$  is known based on the assumed LOS-distribution  $F_{\theta}$ .

Based on this assumed distribution of the observed data  $B_t = b_t$  we set up a Bayesian hierarchical model to estimate the expected number of new ICU admissions based on the data  $B = b = (b_1, \ldots, b_T)$ . Since the number of to be estimated parameters  $\lambda = (\lambda_1, \ldots, \lambda_T)$  is the same as the number of observed data points b, this is close to not identifiable. We therefore impose regularization on the estimated  $\lambda_t$  assuming a smooth change in the daily number of new ICU admissions by setting up a Bayesian hierarchical model

$$\lambda_{1} \sim \operatorname{Prior}(\phi)$$

$$\sigma_{rw} \sim \operatorname{Half-N}(0, 10)$$

$$\lambda_{t} | \lambda_{t-1} \sim N(\lambda_{t-1}, \sigma_{rw}), \ t = 2, \dots, T$$

$$B_{t} | \lambda \sim \operatorname{Pois}(\sum_{k=0}^{K} \lambda_{t-k} \eta_{k})$$

and estimate  $\lambda$  by MCMC sampling.