

1 Observed and unobserved variables

Observed variables

A_t := number of COVID-19 hospital admissions on day t
 C_t := COVID-19 hospital midnight census count on day t

Unobserved (but unambiguously deducible) variables

D_t := number of COVID-19 hospital discharges on day t

Relations

$$\begin{aligned}
 C_t &= \sum_{\tau=0}^t A_{\tau} - \sum_{\tau=0}^t D_{\tau} \\
 D_t &= \sum_{\tau=0}^t D_{\tau} - \sum_{\tau=0}^{t-1} D_{\tau}
 \end{aligned}$$

COVID-19 hospital length of stay distribution: Weibull

$$D_t \sim \text{NegativeBinomial}\left(d_t, d_t + \frac{d_t^2}{\psi}\right)$$

$$d_t = \sum_{\tau=0}^{t-1} A_{\tau} \cdot \pi_{t-\tau}$$

$$\pi_{\tau} := P\left(\begin{array}{c} \text{die or discharged} \\ \text{on the } \tau^{\text{th}} \text{ day} \\ \text{after admission} \end{array} \middle| \begin{array}{c} \text{COVID-19} \\ \text{hospital} \\ \text{admission} \end{array}\right) = \begin{cases} \int_0^{3/2} f_{\text{Wb}}(s; \alpha, \sigma) \, ds, & \text{for } \tau = 1, \\ \int_{\tau-1/2}^{\tau+1/2} f_{\text{Wb}}(s; \alpha, \sigma) \, ds, & \text{for } \tau = 2, 3, \dots \end{cases}$$

Impose prior distribution on, and compute/estimate posterior distribution for:

$$(\alpha, \sigma, \psi) \in (0, \infty) \times (0, \infty) \times (0, \infty).$$

Derivations

$$\begin{aligned}
 f_{\text{Wb}}(s; \alpha, \sigma) &= \left(\begin{array}{c} \text{probability density function of the Weibull distribution} \\ \text{with shape parameter } \alpha > 0, \text{ scale parameter } \sigma > 0 \end{array} \right) \\
 &= \frac{\alpha}{\sigma} \cdot \left(\frac{s}{\sigma}\right)^{\alpha-1} \cdot \exp\left(-\left(\frac{s}{\sigma}\right)^{\alpha}\right), \quad \text{for } s \geq 0
 \end{aligned}$$

$$\begin{aligned}
 F_{\text{Wb}}(s; \alpha, \sigma) &= \left(\begin{array}{c} \text{cumulative distribution function of the Weibull distribution} \\ \text{with shape parameter } \alpha > 0, \text{ scale parameter } \sigma > 0 \end{array} \right) \\
 &= 1 - \exp\left(-\left(s/\sigma\right)^{\alpha}\right), \quad \text{for } s \geq 0
 \end{aligned}$$

Hierarchical Bayesian Model for Hospital Length of Stay

Hence,

$$\begin{aligned}\pi_\tau &:= P\left(\begin{array}{c} \text{die or discharged} \\ \text{on the } \tau^{\text{th}} \text{ day} \\ \text{after admission} \end{array} \middle| \begin{array}{c} \text{COVID-19} \\ \text{hospital} \\ \text{admission} \end{array}\right) = \begin{cases} \int_0^{3/2} f_{\text{Wb}}(s; \alpha, \sigma) \, ds, & \text{for } \tau = 1, \\ \int_{\tau-1/2}^{\tau+1/2} f_{\text{Wb}}(s; \alpha, \sigma) \, ds, & \text{for } \tau = 2, 3, \dots \end{cases} \\ &= \begin{cases} F_{\text{Wb}}\left(\frac{3}{2}; \alpha, \sigma\right), & \text{for } \tau = 1, \\ F_{\text{Wb}}\left(\frac{\tau+1/2}{2}; \alpha, \sigma\right) - F_{\text{Wb}}\left(\frac{\tau-1/2}{2}; \alpha, \sigma\right), & \text{for } \tau = 2, 3, \dots \end{cases} \\ &= \begin{cases} 1 - \exp\left(-\left(\frac{3/2}{\sigma}\right)^\alpha\right), & \text{for } \tau = 1, \\ \exp\left(-\left(\frac{\tau-1/2}{\sigma}\right)^\alpha\right) - \exp\left(-\left(\frac{\tau+1/2}{\sigma}\right)^\alpha\right), & \text{for } \tau = 2, 3, \dots \end{cases}\end{aligned}$$