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

 $D_t := \text{number of COVID-19 hospital discharges on day } t$

Relations

$$C_{t} = \sum_{\tau=0}^{t} A_{\tau} - \sum_{\tau=0}^{t} D_{\tau}$$

$$D_{t} = \sum_{\tau=0}^{t} D_{\tau} - \sum_{\tau=0}^{t-1} D_{\tau}$$

Likelihood assumption #1: Daily death/discharge count ~ Negative Binomial

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

Likelihood assumption #2: Decomposition of d_t by day of admission

$$\begin{array}{ll} d_t &:=& \text{expected number of deaths/discharges on day } t \\ &=& \displaystyle\sum_{\tau=0}^{t-1} \left(\begin{array}{c} \text{expected number of deaths/discharges on day } t \\ \text{among COVID-19 patients admitted on day } 0 \leq \tau < t \end{array} \right) \\ &=& \displaystyle\sum_{\tau=0}^{t-1} \left(\begin{array}{c} \text{number of} \\ \text{admissions} \\ \text{on day } \tau \end{array} \right) \cdot \left(\begin{array}{c} \text{proportion of} \\ \text{death/discharge} \\ \text{after } t - \tau \text{ days} \end{array} \right) \\ &=& \displaystyle\sum_{\tau=0}^{t-1} A_\tau \cdot \pi_{t-\tau} \end{array}$$

Likelihood assumption #3: Random admission-to-discharge delay (i.e. length of stay) ~ Weibull

$$\pi_{\tau} := P \begin{pmatrix} \text{die or discharged} & \text{COVID-19} \\ \text{on the } \tau^{\text{th}} \text{ day} \\ \text{after admission} & \text{admission} \end{pmatrix} = \begin{cases} \int_{0}^{3/2} f_{\text{Wb}}(s; \alpha, \sigma) \, \mathrm{d}s, & \text{for } \tau = 1, \\ \int_{0}^{\tau+1/2} f_{\text{Wb}}(s; \alpha, \sigma) \, \mathrm{d}s, & \text{for } \tau = 2, 3, \dots \end{cases}$$

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

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

Derivations

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

$$\begin{split} F_{\mathrm{Wb}}(\,s\,;\alpha,\sigma) &= \left(\begin{array}{cc} \mathrm{cumulative\ distribution\ function\ of\ the\ Weibull\ distribution} \\ \mathrm{with\ shape\ parameter\ } \alpha>0,\, \mathrm{scale\ parameter\ } \sigma>0 \end{array} \right) \\ &= 1-\exp\!\left(-(s/\sigma)^\alpha\right), \quad \mathrm{for\ } s\geq0 \end{split}$$

Hence,

$$\pi_{\tau} := P \begin{pmatrix} \text{die or discharged} & \text{COVID-19} \\ \text{on the } \tau^{\text{th}} \text{ day} \\ \text{after admission} & \text{admission} \end{pmatrix} = \begin{cases} \int_{0}^{3/2} f_{\text{Wb}}(s; \alpha, \sigma) \, ds, & \text{for } \tau = 1, \\ \int_{\tau - 1/2}^{\tau + 1/2} f_{\text{Wb}}(s; \alpha, \sigma) \, ds, & \text{for } \tau = 2, 3, \dots \end{cases}$$

$$= \begin{cases} F_{\text{Wb}} \left(\frac{3}{2}; \alpha, \sigma \right), & \text{for } \tau = 1, \\ F_{\text{Wb}} \left(\tau + \frac{1}{2}; \alpha, \sigma \right) - F_{\text{Wb}} \left(\tau - \frac{1}{2}; \alpha, \sigma \right), & \text{for } \tau = 2, 3, \dots \end{cases}$$

$$= \begin{cases} 1 - \exp\left(-\left(\frac{3/2}{\sigma}\right)^{\alpha}\right), & \text{for } \tau = 1, \\ \exp\left(-\left(\frac{\tau - 1/2}{\sigma}\right)^{\alpha}\right) - \exp\left(-\left(\frac{\tau + 1/2}{\sigma}\right)^{\alpha}\right), & \text{for } \tau = 2, 3, \dots \end{cases}$$