Importance sampling algorithm

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Data: y^{r, s_{\text{curr}}}, the wILI observations so far; z^{r, s_{\text{curr}}}, a version of y^{r, s_{\text{curr}}} with two extra points estimated from GFT; prior distributions of wILI curves, noise levels, and transformations Result: weighted collection of curves Let \phi(x; \mu, \sigma) be the normal pdf; for a large number of times do Randomly draw f^r, \sigma, \nu, \theta, and \mu from the corresponding priors; Let f^{r, s_{\text{curr}}}(i) = f_4^r(i) = b^r + \frac{\theta^r - b^r}{\max_j f^r(j) - b^r} \left[ f^r \left( \frac{i - \mu^r}{\nu^r} + \arg \max_j f^r(j) \right) - b^r \right]; Calculate weight w = \prod_{i=1}^{\text{length}(z^{r, s_{\text{curr}}})} \phi(z; f^{r, s_{\text{curr}}}(i), \sigma); Let v be a 53-length vector, a possible curve for this season; for i in 1..length(y^{r, s_{\text{curr}}}) do v_i := y_i^{r, s_{\text{curr}}}; end for i in (length(y^{r, s_{\text{curr}}}) + 1)..53 do v_i := f^{r, s_{\text{curr}}}(i); end Add curve v with weight v to the collection of possibilities for this season (the posterior estimate) end
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Algorithm 1: Importance sampling procedure

Note that, since the GFT estimates are not exact, we weight $f^{r, s_{\text{curr}}}$ based on $z^{r, s_{\text{curr}}}$, using both ILINet wILI observations and GFT data, but we construct each possible curve v using $f^{r, s_{\text{curr}}}$ and $y^{r, s_{\text{curr}}}$, using no GFT. However, since ILINet data can undergo revisions, we have also considered versions that construct each v ignoring some of the more recent values in $y^{r, s_{\text{curr}}}$.

To improve computational efficiency, we also use a modified version of the code above that first divides up the possible values of f^r , σ , ν , θ , and μ into bins and estimates the average weight of $f^{r,s_{\text{curr}}}$'s in each bin. By sampling values of f^r , σ , ν , θ , and μ more frequently from the higher-weighted bins (and correcting for this decision in the weight calculation), we are able to construct a collection of curves with a high total weight more quickly than the version above.