Nowcasting Covid-19 onset in the UK

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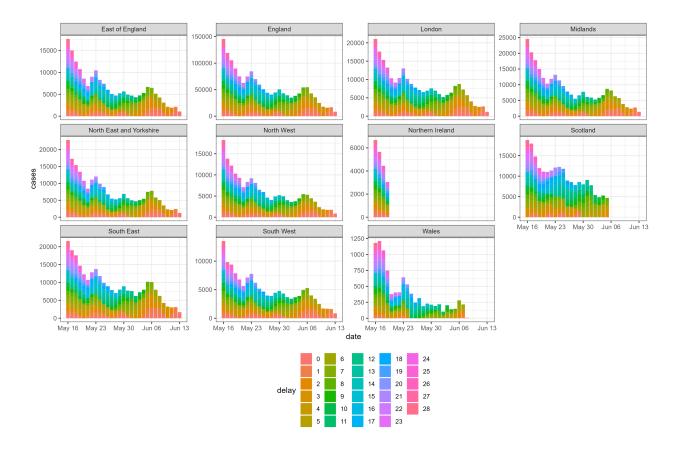
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Software

- Successfully installed epinowcast, managed to run the nowcast for hospitalisations in Germany
- Set up C++ tool chain for CmdStan v2.29.2

Data

- Downloaded newCasesBySpecimenDate from the UK Covid-19 dashboard
 - Aggregated by LTLAs then NHS England regions
 - Age stratification also available newCasesBySpecimenDateAgeDemographics
 - What is a suitable time frame?
- Plotted cases over time by reporting delay
 - Scotland, Wales, and Northern Ireland seem to be working on different reporting schedules
 - What other visualisations are helpful?
 - Would be interesting to see if the distribution of delays changed over time maybe due to testing advice



Theory

The nowcasting problem is described as follows (Höhle and Heiden 2014).

- $n_{t,d}$: number of cases with
 - onset on day t, t = 0, ..., T
 - reported with a delay of d, d = 0, ..., D days, i.e. reports arrive on day t + d.
 - * We may assume that delays can occur only up to a maximum of D days
- We observe $\ltimes = (n_{t,d} : (t,d) \in A_T^m)$ where

$$\{(t,d): \max(T-m,0) \le t \le T, 0 \le d \le \min(D,T-t)\}$$

is the right-angle trapezoidal observation region at time T when using a moving window of size m.

- $N(t,T) = \sum_{d=0}^{\min(T-t,D)} n_{t,d}$: number of cases which occurred on t which are reported until time T
 - The aim of nowcasting is to predict the total number of cases which occurred on t, t = T D, ..., T given the information at time T
- Reporting delay (in days) of a case occurring at time t follows a distribution with probability mass function f_t(d) = p_{t,d}, d = 0, ..., D, where ∑_{d=0}^D p_{t,d} = 1.
 Considering time t and conditioning on number of cases N(t, T) observed by time T, the observed
 - Considering time t and conditioning on number of cases N(t,T) observed by time T, the observed $n'_{t,d}s$ in the reporting triangle due to the right-truncation have a multinomial distribution with size $N(t,\infty)$ and cell probabilities $p'_{t,d} = p_{t,d} / \sum_{i=0}^{\min(D,T-t)} p_{t,i}$ for $d=0,...,\min(D,T-t)$. If we assume that occurrence of new cases follow an underlying inhomogeneous Poisson process,
- If we assume that occurrence of new cases follow an underlying inhomogeneous Poisson process, $N(t,\infty) \sim \text{Poisson}(\lambda_t), t = \max(0,T-m),...,T$, the above conditional formulation of the $n'_{t,d}s$ originating from multinomial sampling can be re-formulated as an unconditional likelihood corresponding to an incomeplete contingency table

$$n_{t,d} \sim \text{Poi}(\mu_{t,d}), \quad (t,d) \in A_T^m, \text{ where } \log(\mu_{t,d}) = \log(\lambda_t) + \log(p_{t,d})$$

• Nowcasting can thus be divided into

- 1. Determining the $\lambda_t's$

2. Determining the $\lambda_t s$ 2. Determining the sequence of $p'_{t,d} s$ 3. Predicting the unobserved $n'_{t,d} s$ to compute the total $N(t,\infty)$ Höhle, Michael, and Matthias an der Heiden. 2014. "Bayesian Nowcasting During the STEC O104:H4 Outbreak in Germany, 2011." *Biometrics* 70 (4): 993–1002. https://doi.org/10.1111/biom.12194.