Modelling the pandemic in Scotland (achievements and lessons)

Christopher J. Banks¹ Ewan Colman¹ Anthony Wood¹ Thomas Doherty² Rowland R. Kao¹



THE UNIVERSITY of EDINBURGH The Royal (Dick) School of Veterinary Studies



13th June 2022

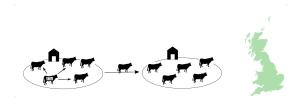
This talk will be on:

- SCoVMod—the Scottish Coronavirus Model;
- use in medium term projections and scenario modelling;
- results from modelling scenarios during the onset of Omicron VoC;
- what we learnt along the way.

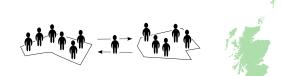
SCoVMod—the Scottish Coronavirus Model

Adapted from an existing agent-based, spatially heterogeneous transmission model:

TBMI:



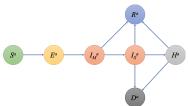
SCoVMod¹:



¹Banks et al., *SCoVMod – a spatially explicit mobility and deprivation adjusted model of first wave COVID-19 transmission dynamics*, Wellcome Open Research (2022)

Chris Banks (Roslin Institute)

Modelling the pandemic in Scotland



- Local transmission:
 - age-structured compartmental model for each location (Census Output Area, 100 people)

- National transmission:
 - Network of locations
 - movements between weighted by census workplace movements
 - and local non-work movements (within IZ, neighbourhood)
 - time-variant weighting according to Google Mobility

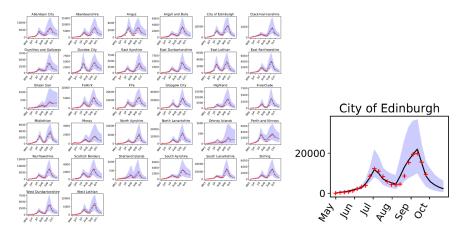
- Deprivation adjustment:
 - Significant correlation between SIMD Health Index
 - Transmission rates adjusted by HI score, per Council Area
- Vaccination:
 - Individuals in the model are flagged as vaccinated according to PHS vaccination data
 - Each vaccination dose has a time-variant efficacy
 - Efficacy data from UKHSA vaccine surveillance

Model fitting

• Case ascertainment:

- A separate model² estimates case ascertainment
- and therefore infers the number of daily infections
- from PHS case data and the ONS Infection Survey
- $\bullet\,$ Case ascertainment in Scotland has been roughly $\,25\%\,$
- Parameter estimation:
 - Fit against inferred infections (early versions used NRS deaths)
 - ABC-SMC, 9 parameters
 - Amazon EC2 (300 machines x 8 cores)
- Time-variant local transmission rates
 - Post-ABC, piece-wise, least-squares fit
 - inflection points (largely) correspond to NPI events
 - per Council Area

²Colman et al., *Estimating the proportion of SARS-CoV-2 infections ascertained through diagnostic testing*, medRxiv (2021)



Aim: take the existing model fit for Delta and project likely scenarios for the Omicron outbreak

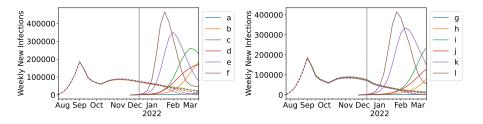
- Two-strain variant of the model constructed
- Complete cross-immunity between strains was assumed
- We vary transmission rate and vaccine efficacy for Omicron

		Efficacy multiplier			
		2 dose	+15 weeks	Booster	
Vaccine Escape Level	1	1	1	1	
	2	1	0.45	0.8	
	3	0.72	0.15	0.63	

 $\bullet~$ UKHSA estimate of between 2.25 \times and 3.2 \times transmission advantage

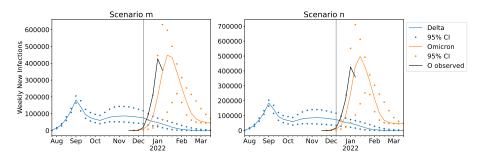
Omicron VoC scenarios

Vaccine Escape Level					
		1	2	3	
Transmission Level	1	a	b	с	
	2	d	e	f	
	1+NPI	g	\mathbf{h}	i	
	2+NPI	j	k	1	



Omicron VoC adjusted scenario

- Best-fit scenario taken (high TR advantage)
- Growth rate adjusted to fit observed SGTF cases



- to give a TR advantage of between 3.6× and 4.0× (higher/lower vacc. esc. resp.)
- Timing a bit off, but magnitude predicted fairly well!

- Having a ready-to-go model useful for scenarios and informing policy
- Weekly modelling outputs to Scottish Government
- Ability to quick make use of privileged access to spatially detailed PHS data
- Capturing complex spatial heterogeneity (in disease and mobility)
- Making well-founded projections when it mattered (omicron outbreak)
- Doing it with a small team!

Lessons? (A really big slide with lots of stuff!)

- Being prepared: having a (almost) ready-to-go model was key
- Having the people to do it:
 - With the availability to work flexibly and funded appropriately
 - Research Software Engineer (Tom)
 - Core Scientist (me)
 - willing funders (Ewan, Wellcome Trust)
- RSE was key, with his knowledge of good software engineering:
 - Test Driven Development: unit tests so new code didn't break old code, large suite of tests built up
 - Ideally code review/paired programming, but we only had one RSE!
 - One RSE being a bottleneck we shifted some of the adaptation to data prep, to share the load:

e.g. movements treated as in TBMI, but with an external movement generator based on data $% \left({{{\rm{BMI}}} \right)$

- Hard to publish: useful and relevant work, but not ground-breaking
- Fitting a time-variant, spatially heterogeneous, event-driven transmission rate is hard and labour intensive.

Chris Banks (Roslin Institute)

Modelling the pandemic in Scotland

Roslin Institute:

Rowland Kao Ewan Colman Anthony Wood Aeron Sanchez Gianluigi Rossi Paul Bessel Anne-Sophie Ruget Daniel Balaz

Others:

Tom Doherty (Strathclyde) Jess Enright (Glasgow) Gael Beaunée (INRAE) Gavrila Puspitarani (Vetmeduni Vienna) Adam Kleczkowski (Strathclyde) Katie Atkins (Edinburgh) Oliver Tearne (APHA) Mark Arnold (APHA)