

Scottish COVID-19 Response Consortium

Monday, 27th July 2020



Network Simulation



https://github.com/ScottishCovidResponse/simple_network_sim



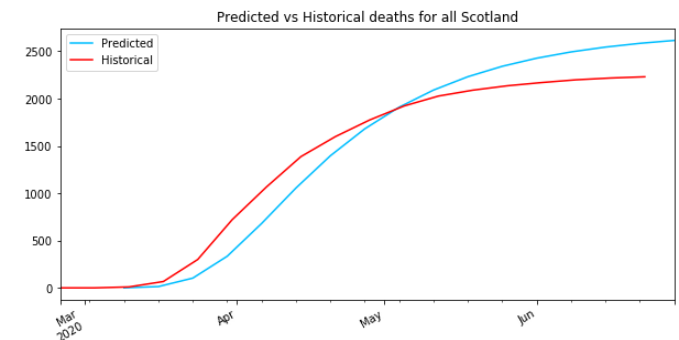
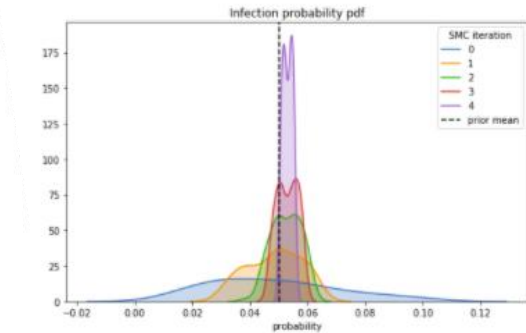
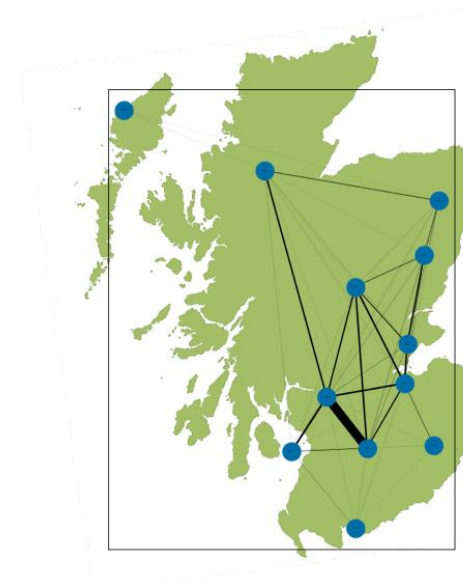
Jess Enright (Glasgow)
Bob Turner (Sheffield)
Rafael de Almeida (Man Group)
Chris Hughes (Man Group)
Fedor Gorokhovik (Man Group)

Will Pettersen (Glasgow)
 Anthony O'Hare (Stirling)
 Leighton Pritchard (Strathclyde)
 Alex Konovalov (St Andrews)
 Jason Dykes (City)
 Jo Wood (City)

Language	Python
Spatial scale	Data dependent – currently Health Board or IZ
Model type	Spatially-embedded network of age-structured populations, deterministic or stochastic, compartmental model
Features	<ul style="list-style-type: none"> - Movement flows from data or simulated - Age-class mixing from input matrix - Local and long-distance spread - Dynamic movement and contact damping
Mode	Forward simulation, daily timesteps Basic inference implemented – ABC-SMC

Key questions

- Control related to age class or movements
- Importance of network flows, impact of changes to flows



Network Simulation



https://github.com/ScottishCovidResponse/simple_network_sim

Significant achievements over the RAMP period

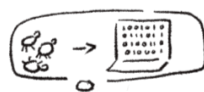
Software engineering	Many major improvements in design, automated testing, documentation, reproducibility, and more!
Modelling functionality	<ul style="list-style-type: none">- Choice to focus on network-of-populations- Generalisation to arbitrary regions given data- Generalised compartmental model- Time-varying activity damping- Stochastic model operation- Generating realistic behaviour- Enormously improved software choices!
Data Interface	<ul style="list-style-type: none">- Move to inputs from files instead of hard-coded or simulated- Using real data as input- Data Pipeline API integration
Inference and Sensitivity	<ul style="list-style-type: none">- Linear and non-linear sensitivity analysis- Implementation of ABC-SMC for basic inference



Contact Tracing

Sibylle Mohr, Louise Matthews
(EPIC / Glasgow)

Sam Brett, Vino Mano (Man Group),
John Nonweiler (UKAEA), Ed Townsend



<https://github.com/ScottishCovidResponse/Contact-Tracing-Model>

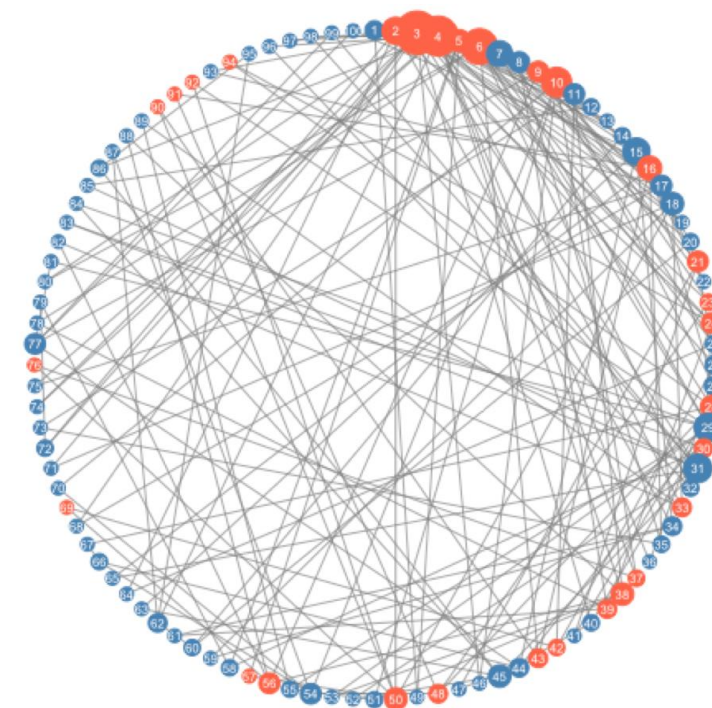


Language	Java
Spatial scale	No explicit spatial scale but a form of 'location' is retained
Model type	Individual-based stochastic network model <ul style="list-style-type: none">• Age-structured individuals• Weighted contacts
Key Feature	Flexible contact tracing scenarios
Mode	Forward simulation, user-defined timesteps (quarter of day)

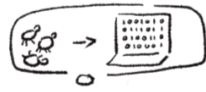
Key questions

What are the optimal contact tracing protocols?

- Benefits of blanket vs tailored actions and advice for alertees?
- How can we reduce the burden of contact tracing whilst maintaining high levels of efficacy
 - Targeted alerts / rapid testing / release of contacts from isolation
- How much more effective does the tracing become if tracing of secondary contacts is allowed?
- What is the impact of basing alerting on a positive test rather than symptoms?



Contact Tracing: Key features



Sibylle Mohr, Louise Matthews
(EPIC / Glasgow)



Sam Brett, **Vino Mano** (Man Group),
John Nonweiler (UKAEA), Ed Townsend

Explore contact-tracing policies

Isolation measures

- Blanket 14-day isolation vs. dependent on the time of contact (0-14 days)

Alerts / testing

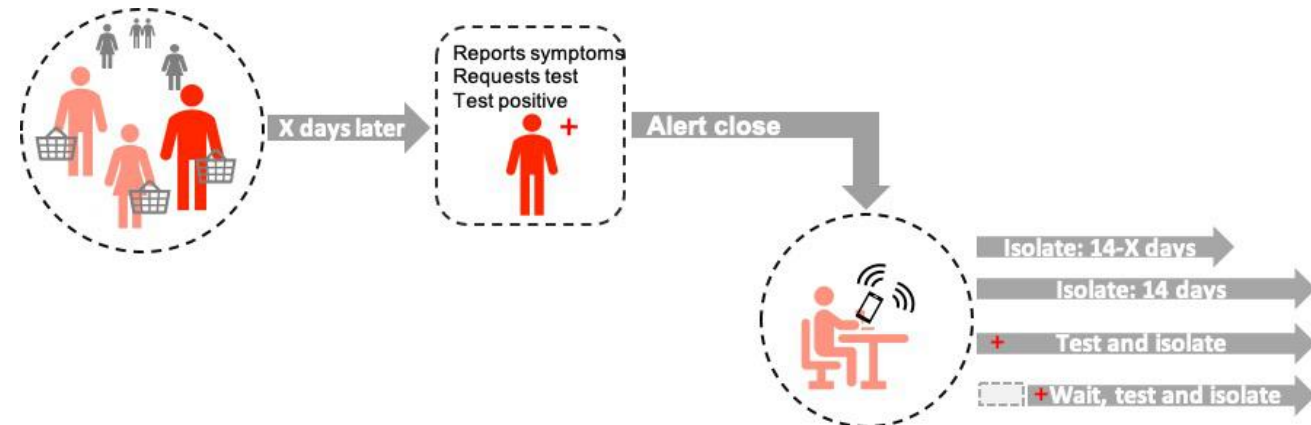
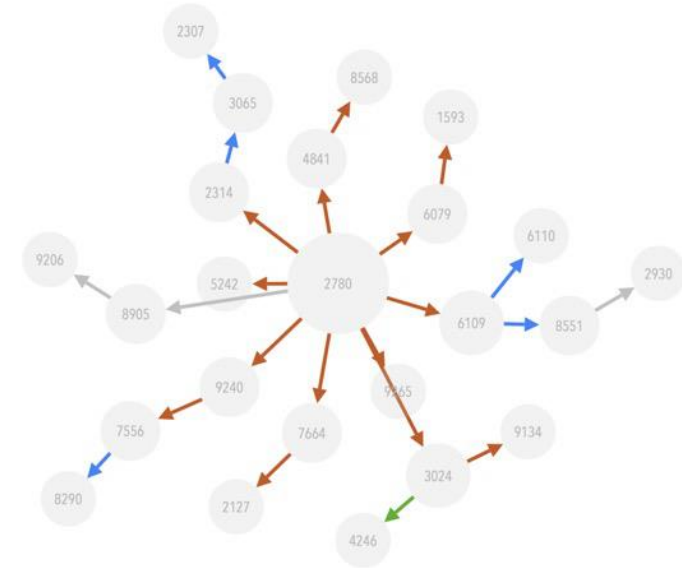
- targeting of alerts at individuals **most likely** to be infected
- testing of individuals to **release** their contacts from isolation

Different 'alerting' scenarios

- Alerting based on **symptomatics**
- Alerting based on **positive test**

Contact tracing

- Direct contacts only
- Also **secondary** contacts (contacts of direct contacts)



Contact Tracing Team



Major RAMP Achievements

Software development

- Code clean-up; codebase tested / integrated / optimised; re-factoring; added software tests / unit tests...
- Documentation, verification, integration into the SCRC environment...

Model functionality

- Multiple, sophisticated contact-tracing scenarios / secondary contacts / global lockdown policies / testing scenarios / test efficacy
- Model runs on various contact data, together with list of age-structured individuals; Random infection / targeted seeding
- Complex statistics output (R calculations; traced / isolated / infected...)

Inference

- Sensitivity analysis for model parameters / policy parameters
- Comparison between CTM model and ODE model

Visualisation

- Multiple static and dynamic visualisation of different output data
 - Network data / compartmental outputs / infection maps

And so much more!!

- Data API for Java is taking shape; new documentation / inferences / visualisation...

RSE core team	Sam Brett Vino Mano (Man Group) John Nonweiler (UKAEA) Edward Townsend
Inference / model assessment	Rikiya Takahashi Fedor Gorokhovik (Man Group)
Visualisation team	Cagatay Turkay (Warwick) Dan Archambault (Swansea) Kai Xu (Middlesex)
Contact tracing options	Louise Matthews (UoG) Jamie Prentice (UoG)

Spatial Simulation

Claire Harris (EPIC / BioSS)

James Cook (UKAEA)



<https://github.com/ScottishCovidResponse/Simulation.jl>



Language	Julia
Spatial scale	1 km ² grid, Scotland / UK
Model type	Explicitly spatial, stochastic, compartmental model
Features	<ul style="list-style-type: none">- Airborne and environmental spread of virus- Local spread vs. long-distance commuting- Impact of climate / environment on transmission- Demographic structure, population density- Inference and visualisation frameworks
Mode	Forward simulation, daily timesteps

Key questions

- What effect does long term pollution have on susceptibility / severity of COVID-19?
- What is the effect of weather/environment on the virus?
- What are the relative contributions of direct/indirect transmission?



RSE team: Alex Robson, Bella Wu, Eric Perim Martins and Sean Lovett (Invenia)

Epidemiology: Glenn Marion and Iain McKendrick (BioSS), Louise Matthews and Richard Reeve (Glasgow)

Inference: Ben Swallow and Simon Babayan (Glasgow)

Visualisation: Aidan Slingsby (London City)

Data: Sonia Mitchell and Ciaran McMonagle (Glasgow), Stephen Catterall (BioSS) and Jeremy Walton (Met Office)

Spatial Simulation

Claire Harris (EPIC / BioSS)

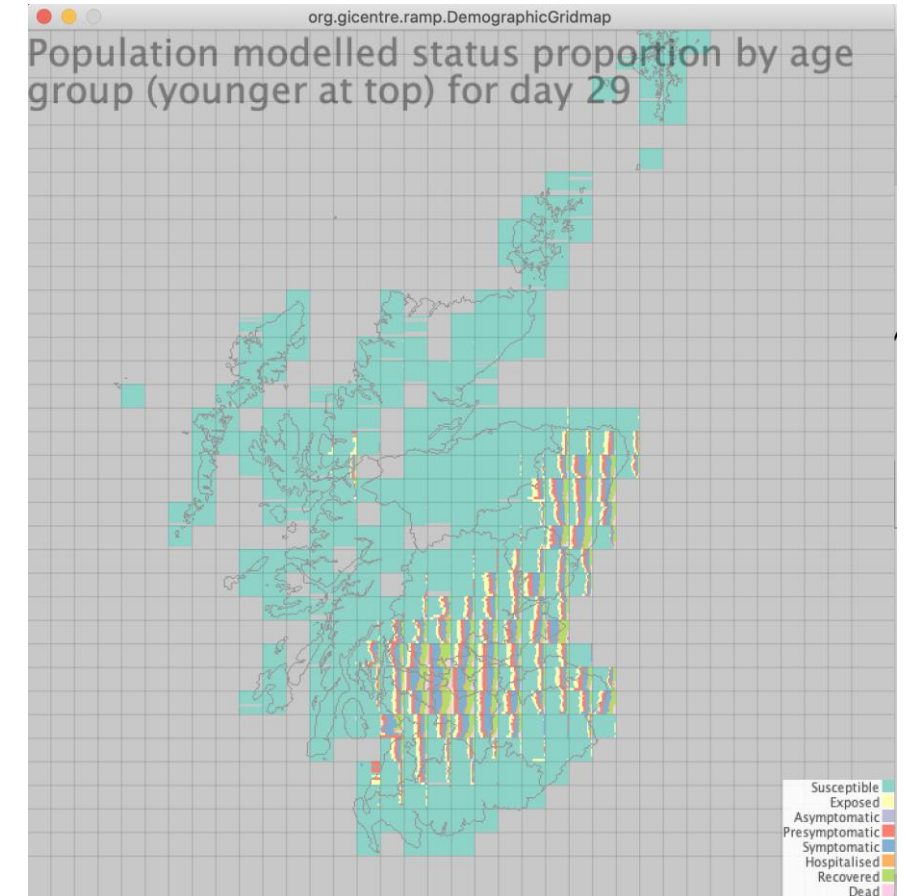
James Cook (UKAEA)

 <https://github.com/ScottishCovidResponse/Simulation.jl>



Major developments

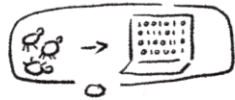
- Codebase:
 - Integrated, tested, documented, optimised
 - Python wrapper for data pipeline
 - Fully deterministic version of code
- Epidemiology:
 - Mostly everything!
- Data:
 - Data flowing from the pipeline
 - Demographics/Epi params/Weather/Pollution
- Inference:
 - ABC-SMC interface in R that can recover initial parameters
- Visualisation:
 - Fully interactive Java visualisation, with multiple modes, scaling and spatiotemporal dynamics



Individual-based Simulation

Paul Bessell (Edinburgh),

John Nonweiler (UKAEA),
Blair Archibald (Glasgow),
Bob Taylor



https://github.com/ScottishCovidResponse/Covid_Simulation_Model

Language	Java
Spatial scale	Defined area (eg Lothian HB, Orkney)
Model type	Individual-based stochastic simulation
Features	<ul style="list-style-type: none">- Individual hourly movements and mixing- Implicit network structure- Full age and household structure- Mixing and transmission in distinct community compartments- Health boards and care homes
Mode	Forward simulation, hour time steps
Fitting	Relies on parameters and visual comparison to data
Data	Most key parameters are now evidenced, but would be good to complete the list

Model purposes

- Detailed simulation of different communities
- Completely flexible reopening/reclosing of businesses and compartments of society
- Comparison between different communities / areas
- Scenarios for opening schools
- Flexible addition of new societal compartments (could add theatres, cinemas etc)
- Shielding populations
- Strategies for roll-out of vaccination or new tests
- Role of superspreaders
- Informing other models with network structure

Individual-based Simulation

Major RAMP Achievements

Software development

- Simple terribly written code has been transformed into a very efficient and complete epidemiological model
- Complete testing environment with the model
- 1 run of a 200 day simulation of 910,000 population on DiRaC is ~22 mins
- Full software documentation

Model functionality

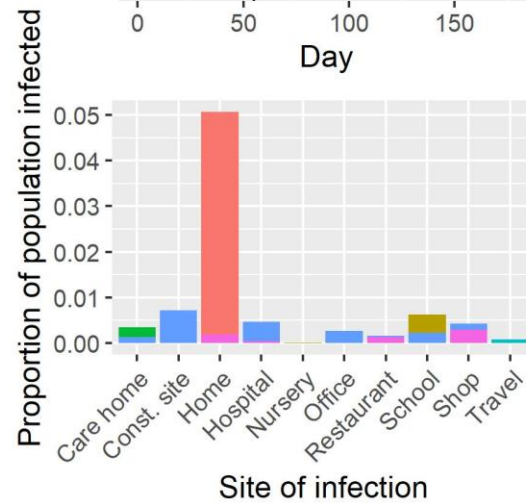
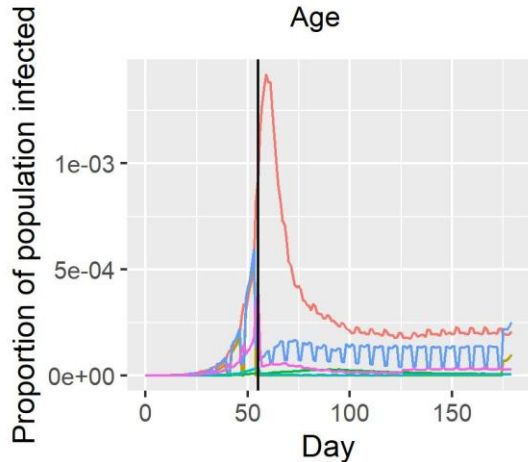
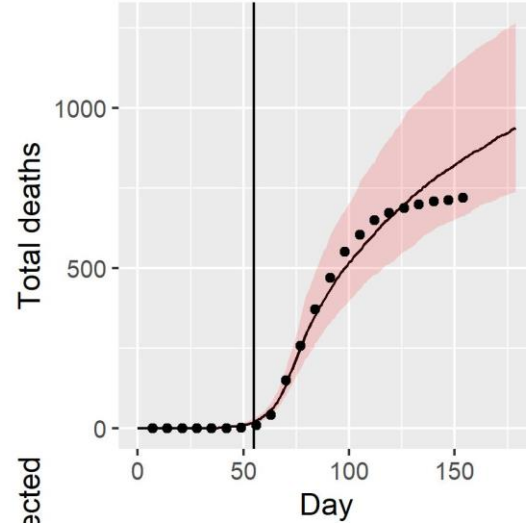
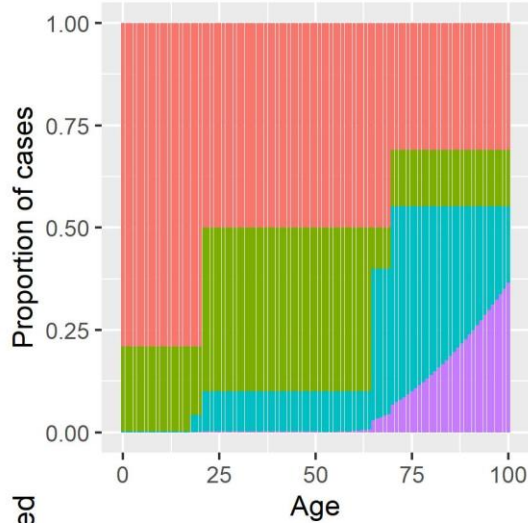
- Fully age structured disease progression
- Age structured household structure and implicit contact network
- Fully age-structured non-Covid hospital utilisation
- Disease transmission during travel
- Build-your-own lockdown release scenarios

Fitting

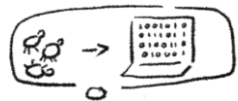
- (For Lothian at least) Model broadly fits the mortality, hospitalisation and R values
- Will run it through Thibaud's model

Further steps

- Write up the epidemiological documentation
- Get better contact parameter data



ABC-smc



Thibaud Porphyre (Edinburgh)



Peter Fox (UKAEA)



https://github.com/ScottishCovidResponse/Covid19_EERAModel



Language	C++
Spatial scale	Flexible
Model type	Age-structured compartmental stochastic model
Features	<ul style="list-style-type: none">- Different transmission behaviour between community and HCW- Impact of bed capacity on hospitalisation rate
Mode	Inference – ABC-smc Forward simulation



RSE team: Sanket Gadgil, Kristian Zarębski, Qingfeng Xia (UKAEA), Andy Bennett



Epidemiology: Mark Bronsvoort, Paul Bessell (Edinburgh), Louise Matthews (Glasgow)



Uncertainty quantification: Ben Swallow (Glasgow), Ian Vernon (Durham), and the rest of #UQ team



Data: Stella Mazeri (Edinburgh)



Model purposes

- Inference of key epidemiological parameters
- Impact of scales (national vs regional vs health board)
- Prediction of community level infection
- Quantification of number of primary infection
- Optimization lifting movement restrictions strategies

ABC-smc

Major RAMP Achievements

Software development

- Simple terribly written code has been transformed into a very efficient and complete epidemiological model
- Complete testing environment with the model
- Full software documentation
- Full epidemiological documentation
- Sensitivity analysis

Model functionality

- Both inference and forward production implemented
- Fully age structured disease progression
- Role of HCW in the case reporting

Fitting

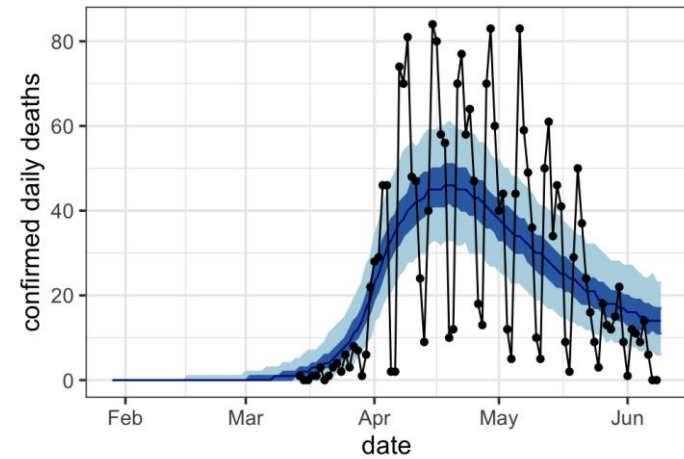
- Model broadly fits the mortality, hospitalisation and R values for Scotland
- Inference framework validated

Further steps

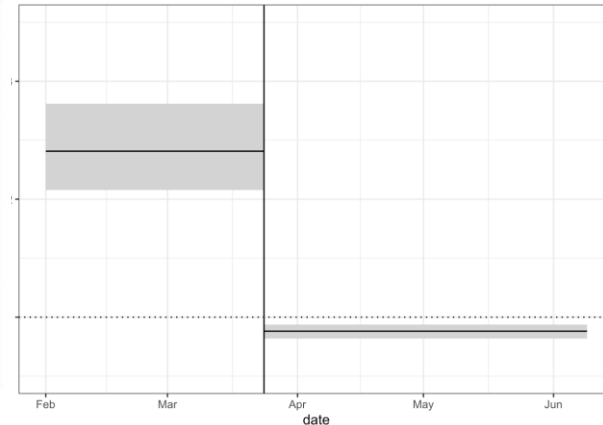
- Uncertainty quantification
- Coding and assessing lockdown strategies
- Fit on all health board in UK



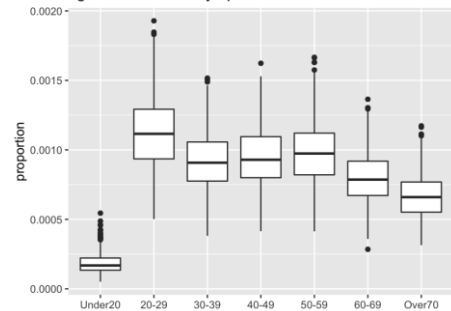
death: Scotland



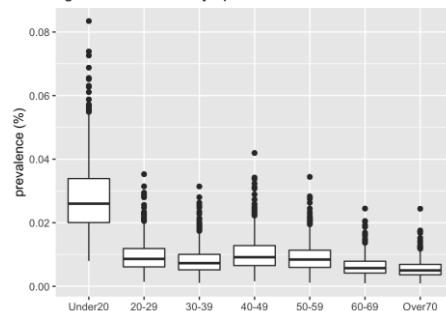
$u=0$; $\theta_{r}=11$ days



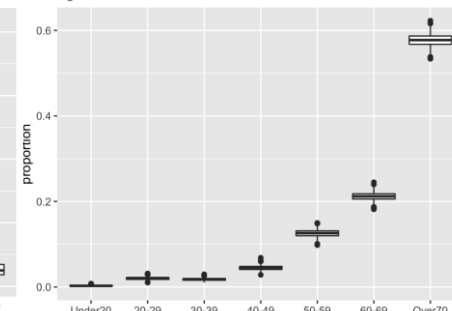
age distribution of symp infectious



age distribution of asymptomatics+recovered in communit



age distribution of deaths



BEEPmbp

Chris Pooley (BioSS / EPIC)
Glenn Marion (BioSS/ EPIC)



<https://github.com/ScottishCovidResponse/CoronaPMCMC>



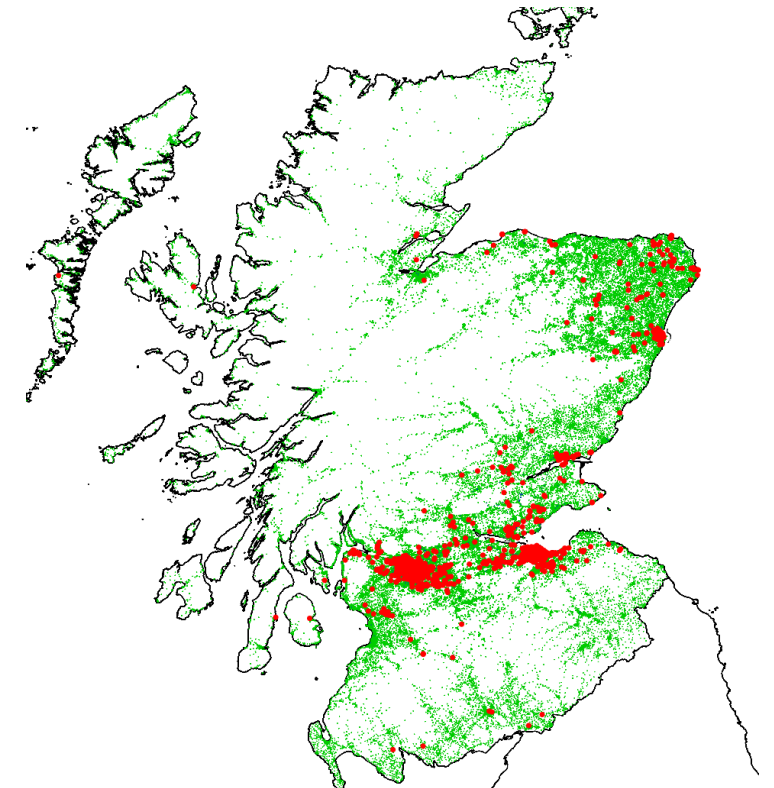
Ian Hinder (Manchester)
Robin Williams (Bristol)

Data: Richard Bailey (Roslin)
Stephen Caterall (BioSS)
Model: Glenn Marion (BioSS)
Andrea Doeschl-Wilson (Edinburgh)

Language	C++
Spatial scale	Middle Layer Super Output Areas (MSOA) or output areas (OA), Scotland
Model type	Individual-based, spatial, stochastic compartmental model
Features	<ul style="list-style-type: none">- Individual risk factors, e.g. age, gender- Area risk factors, e.g. population density- Time variation in transmission- Calculation of model evidence (ME)- Optimized parallel implementation
Mode	Inference using multi-temperature MCMC with model based proposals

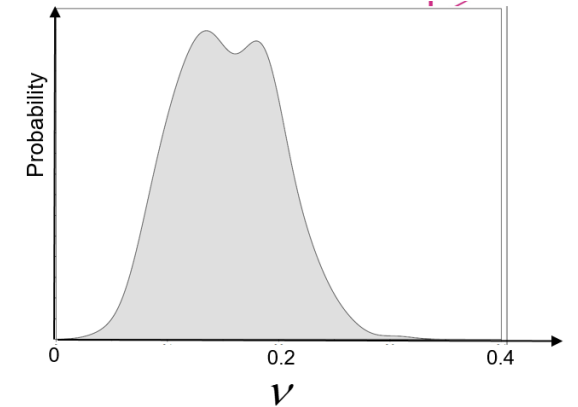
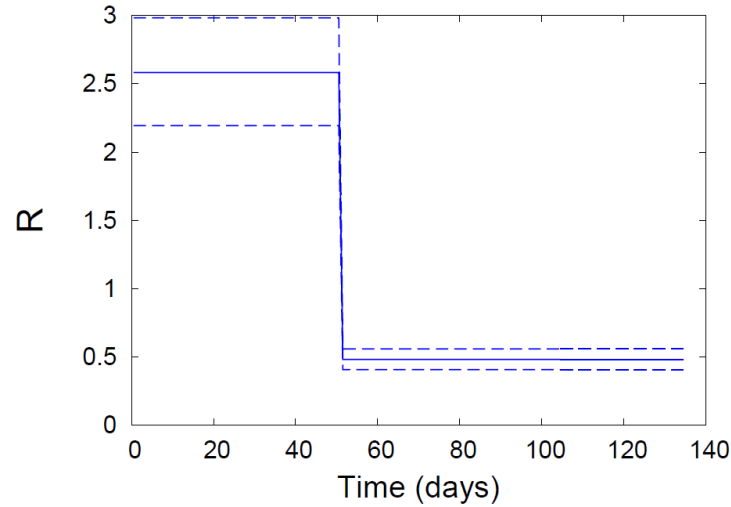
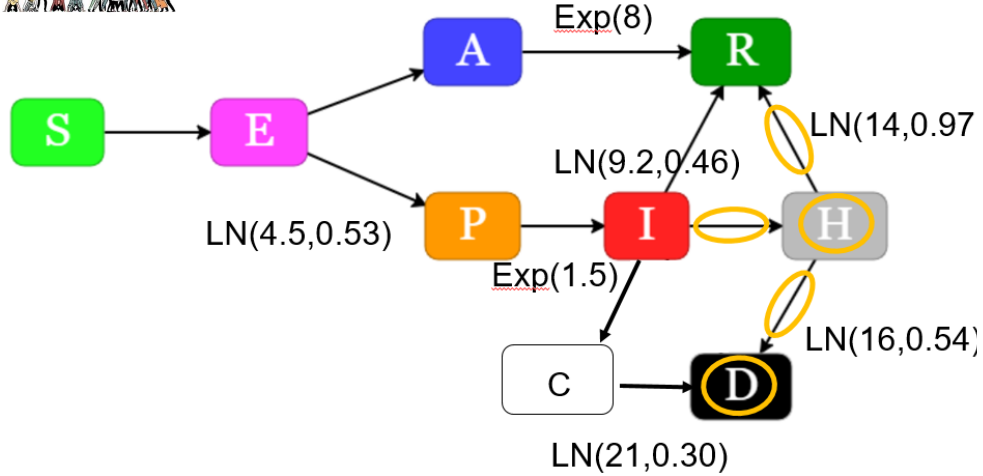
Key questions

- How is R changing as a function of time?
- What determines the spatial spread of disease?
- What are the effects of government policies?





BEEPmbp



$$R_a = \bar{R} \left(\frac{d_a}{\bar{d}} \right)^\nu$$

$\nu = 0$ Frequency dependent
 $\nu = 1$ Density dependent

Model parameter inference

- Data: National hospital admissions /populations and regional death data.
- Distributions for certain transitions.

To Do...

- Comprehensive validation.
- Results for age structured populations with sex.
- Incorporation into data pipeline.

Variation in R

- Captured time variation in disease transmission before/after lockdown.

Population density

- Higher disease transmission in areas of high population density

- Different area factors, e.g. deprivation, pollution.
- Sensitivity analysis.
- Use model evidence to identify best available model for Covid-19 in Scotland.

Open Epidemiology



Requirements

- Reproducible
- Traceable
- Validated
- Open source
- Open data

Open Epidemiology



<https://github.com/ScottishCovidResponse>

Responsibility

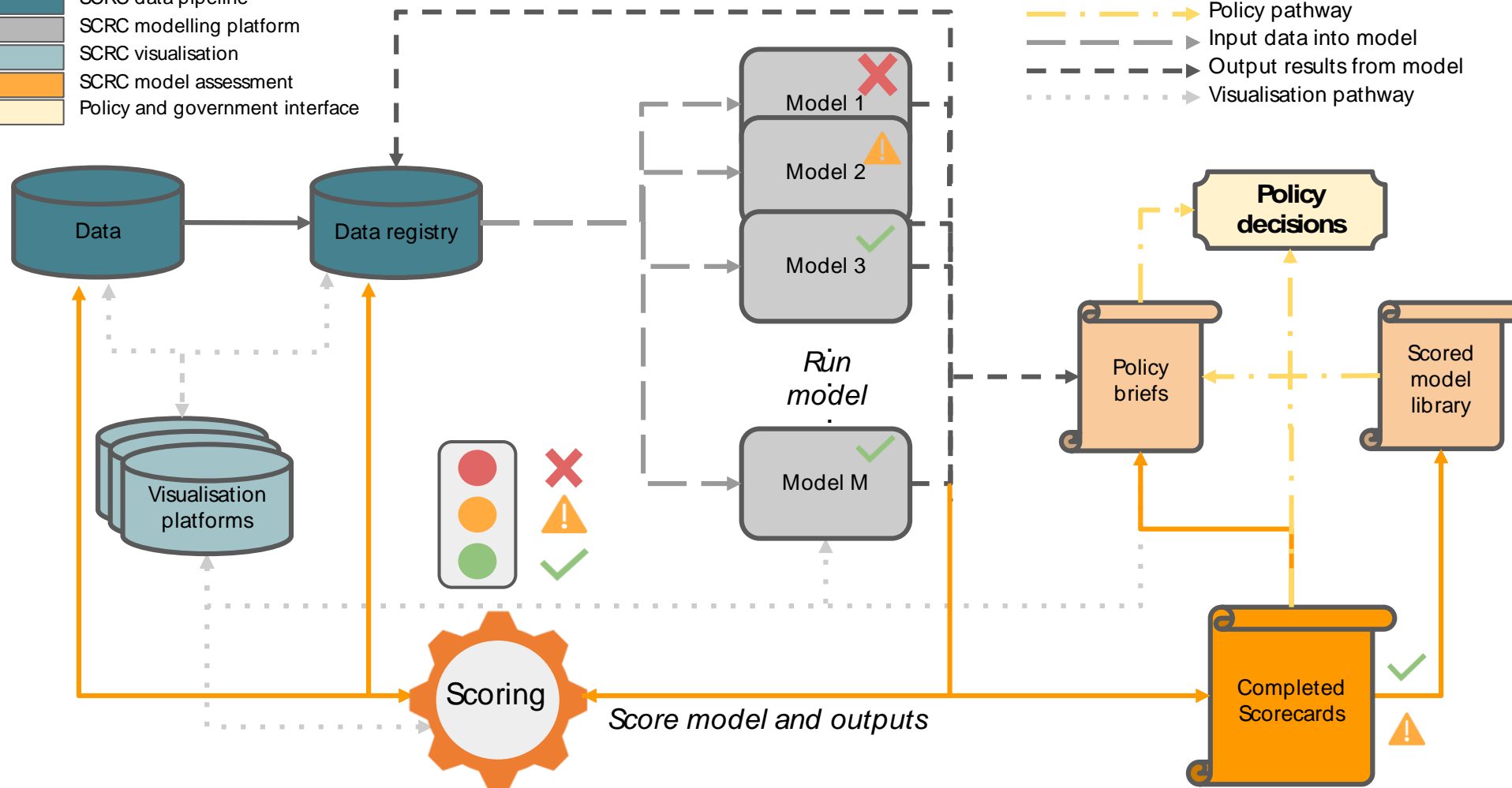
	SCRC data pipeline
	SCRC modelling platform
	SCRC visualisation
	SCRC model assessment
	Policy and government interface

Action

	Modelling platform assessment
	Policy pathway
	Input data into model
	Output results from model
	Visualisation pathway

Requirements

- Reproducible
- Traceable
- Validated
- Open source
- Open data



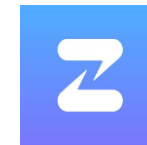
RSE

Alys Brett (UKAEA)

Team processes: Lead RSEs

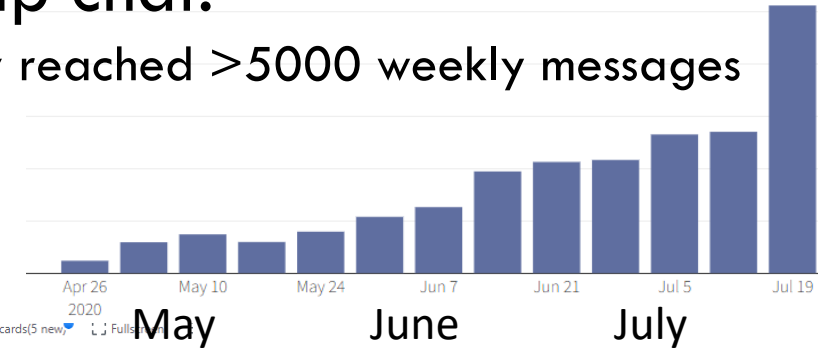
Collaboration infrastructure: Andrew Lahiff, Jonathan Hollocombe, Nathan Cummings (UKAEA). Collaboration: everyone!

Enabling software teamwork



Zulip chat:

Now reached >5000 weekly messages



Github

33 repositories, 14 teams, 13 projects

Simulation.jl development
Updated 3 days ago

The Kanban board shows the following columns and tasks:

- To Do (22 items):**
 - Integrate benchmarking into travis (Quality, RSE, Simulation.jl)
 - Write outputs as a raster (RSE, Simulation.jl, Starter Issue)
 - Make first steps towards reproducibility (Domain Expert, RSE, Simulation.jl)
 - Small improvements to readfile functions (RSE, Simulation.jl, Starter Issue)
 - Advanced testing comparison to deterministic ODEs (RSE, Simulation.jl)
- In Progress (1 item):**
 - Age-related mixing (Domain Expert, Simulation.jl)
- Review in progress (0 items)**
- Reviewer approved (1 item):**
 - WIP: Jc/deterministic: Enable Simulation.jl to run as a deterministic code rather than stochastic (Changes approved)
- Done (63 items):**
 - Spatial mixing (Domain Expert, Simulation.jl)
 - Update documentation (Changes approved)
 - PyCall shim for data_pipeline_api (Simulation.jl, data pipeline api)
 - Deterministic models and testing (Domain Expert, Functionality, RSE, Simulation.jl)

RSE

Alys Brett (UKAEA)

Software Quality

- Improving code clarity and software structure
- Automated testing and code analysis



COVID-19 EERA Model

This is the documentation website for the EERA COVID EERA Model repository on GitHub.

Documentation

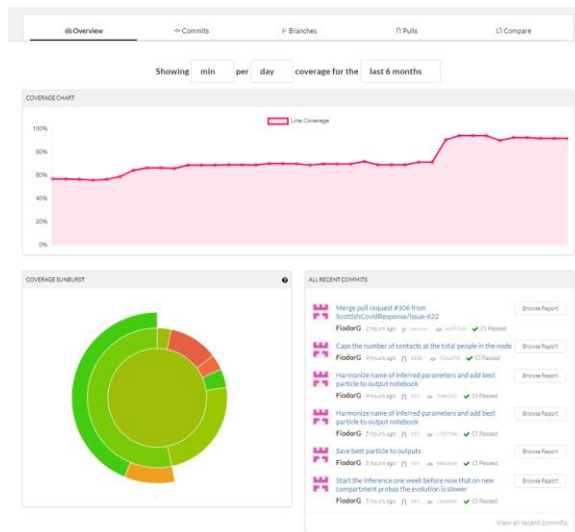
[Model Overview](#) [Doxygen Code Documentation](#)

Code Check Reports

[Flawfinder](#) [Clang Tidy](#) [Code Coverage](#) [CPP Check](#) [Sim C++](#)

Builds

Branch	Status
master	Covid19EERAModel no status
dev	Covid19EERAModel passing

A screenshot of the Travis CI dashboard. The top navigation bar includes "Dashboard", "Changelog", "Documentation", and "Help". The main content area shows the repository "ScottishCovidResponse / simple_network_sim" with a "build passing" status. Below the repository name, there are tabs for "Current", "Branches", "Build History", and "Pull Requests". The "Branches" tab is active, showing a table of branches. The table has columns for "Default Branch", "Active Branches", "# passed", "time ago", and "author".

Default Branch	# passed	time ago	author	status
✓ master 72 builds	# 796 passed 2 hours ago	d2f57c0	FiodorG	✓ ✓ ✓ ✓ ✓
Active Branches	# passed	time ago	author	status
✓ issue-622 3 builds	# 794 passed 4 hours ago	f24a670	FiodorG	✓ ✗ ✓

RSE

Alys Brett (UKAEA)



Improving reproducibility and trust

- Data pipeline integration
- Managing configuration, randomness, build, release run processes

Software Checklist

https://github.com/ScottishCovidResponse/SCRC-internal/blob/master/software_checklist.md

- Framework for evaluating software engineering aspects of model readiness
- Completed for current status of all models

RSE

Alys Brett (UKAEA)

Bob Turner
(Sheffield)

James Cook
(UKAEA)

Simple Network Sim

- **Synchronised docs and code** using docstrings & typehints to making it easy to get into the codebase
- **Extensive automation of unit and regression tests and code-checking through CI** - lets us focus on adding features and fixing bugs
- Adopted **industry standard Python data science tools** `pandas` and `numpy` - peer-assured variable-types and functions
- **Tight control of execution using `conda` virtual environment** to help ensure reproducibility
- **Great multidisciplinary working** and input to current code from a range of contributors: academics, quantitative researchers, postdocs, engineers

Simulation.jl

- **Software teamwork practices (first multi-developer team)**
 - Issue boards & labelling, contribution and review guidelines etc
- **Automated testing**
 - Travis & Appveyor for CI
 - Codecov for test coverage
- **Extensive testing of new epidemiology code**
- **Reproducibility improvements**
 - "Deterministic hijacking" of the stochastic model to allow regression testing
 - Run scripts, TOML files for code & dependency versions
- **Self-contained example run scripts** serve as end to end tests
- **Integration of data pipeline**

RSE

Alys Brett (UKAEA)

Contact Tracing Model and Covid Simulation Model

- Many **model refinements and new features**
- Initial **code clean-up** then incremental **improvements to design & implementation**
- **Automated tests and CI**
 - 90% unit test coverage, regression tests, basic automated epidemiological/model-validation tests
- **Documentation** for user, developers & to understand algorithms
- **Improved reproducibility and trustworthy results**
 - refactoring RNG usage
 - moving parameters to an input file, adding logging of git hash and inputs
 - R scripts for validation, version numbering and release processes
- Enabled **parallel runs** of models on DiRAC HPC
- Implementing the Java **data pipeline**
 - Close to end-to-end model run (using CSV instead of HDF5 files initially)

John Nonweiler (UKAEA)
Sam Brett

RSE



Alys Brett (UKAEA)

EERA Model



Peter Fox (UKAEA)

New functionality: Multiple model structures, forward prediction / parameter inference modes

Reproducibility: Parameterised config of random number generation, Logging inputs, outputs, versions


Software engineering and code quality

- Extensive **refactoring and architecture improvements**
- CMake **build system**
- Unit and regression **test frameworks**
- **CI automation** of: tests, documentation generation (Doxygen), code coverage, static analysis

- Created **project dashboard**

https://scottishcovidresponse.github.io/Covid19_EERAModel

BEEPmbp



Ian Hinder (Manchester)

- Development of C++ interface to Python **data pipeline** code, and integration with model (in progress)
- Fixed configuration in header files replaced with **TOML configuration file**
- Storing code **provenance** information and making runs **reproducible**
- Improved **code structure**, i.e. making it more modular
- Implemented CI (**automated testing**), as well as **regression testing** (flag whether a code change affects the numerical results)
- Added **automated source code analysis** to catch bugs - found several

RSE – External Models



Alys Brett (UKAEA)

Imperial CovidSim Model



Peter Hill (York)

Integration with SCRC data pipeline

- Via a Python wrapper layer using pipeline API
- No need to touch original code
- Can treat chosen version as a black box whilst verifying inputs & outputs

LSHTM Model



Kristian Zarebski (UKAEA)

- **Rationalised parameter initialisation**
 - external config at start of run
- **Faster runs** with single scenario selected via parameter
- **Integration with the SCRC data pipeline**
 - R frontend for python to interact with pipeline API
 - Inputs converted to HDF5 & TOML - now moving to data pipeline locations
- **Collaboration with Swansea RSEs**

Data

Ruth Dundas (UoG), Iain McKendrick (BioSS)



Initial specification of Data Registry: with Aaron Reeves (SRUC)

Specification of data and metadata workflows

Literature review: procedure and associated data extraction tool

Dataset assessment: 'intrinsic quality', 'representativeness' and 'fitness for purpose' assessment procedure

Meta-analysis and distribution analysis: Jim Lewsey, Houra Haghpanahan, Ryan Field, Janine Illian, Caitlin Fox, Mahi Siddika, Megan Laxton, Martje Rave

Data brokering: Richard Bailey, Ciaran McMonagle

Specialist Dataset expertise: Ruth, Ciaran, Jeremy Walton, Denise Brown, Stephen Catterall, Amira Elayouty

Data source and provenance IDing and wrangling: Sonia Mitchell, Ruth, Ciaran, Denise

Data team: Richard, Ciaran, Sonia, Ryan, Jeremy Gribble, Hannah Jevans, Nardus Mollentze, Denise

Data pipeline



<https://github.com/ScottishCovidResponse>

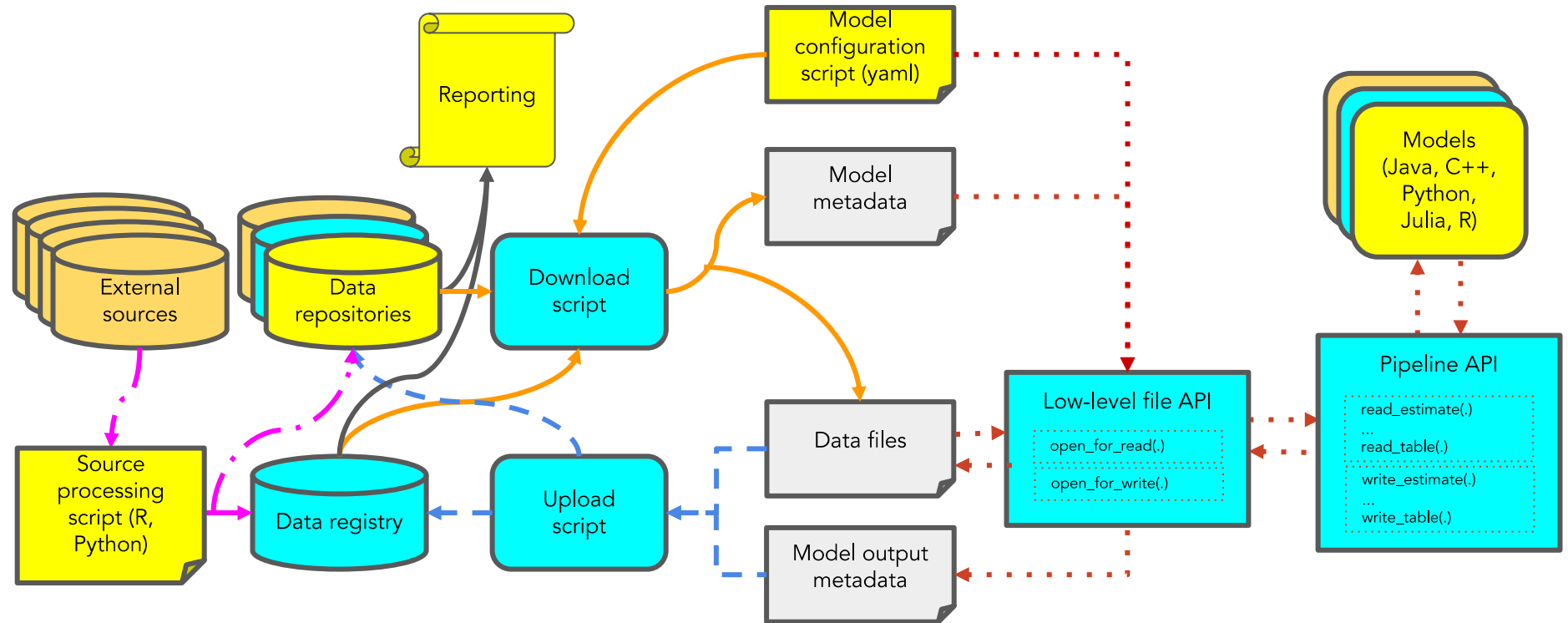


Requirements

- Reproducible
- Traceable
- Open source
- Open data

API

- C/C++
- Java
- Python
- Julia
- R



Data and metadata flow

- - - - - Enter source data into pipeline
- - - - - Download inputs to model
- - - - - Run model using Data Pipeline API
- - - - - Upload outputs to pipeline
- - - - - Reporting on model outputs

Component

- Provided by user
- Core pipeline / SCRC
- From other source
- Generated by pipeline

Data pipeline for input



<https://github.com/ScottishCovidResponse/SCRCdataAPI>

<https://github.com/ScottishCovidResponse/SCRCdata>



Sonia Mitchell
Ciaran McMonagle
Nardus Mollentze
Ryan Field



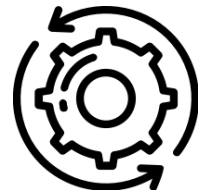
> 31,000 papers uploaded to the registry



8 data products uploaded to the registry



4 upload templates



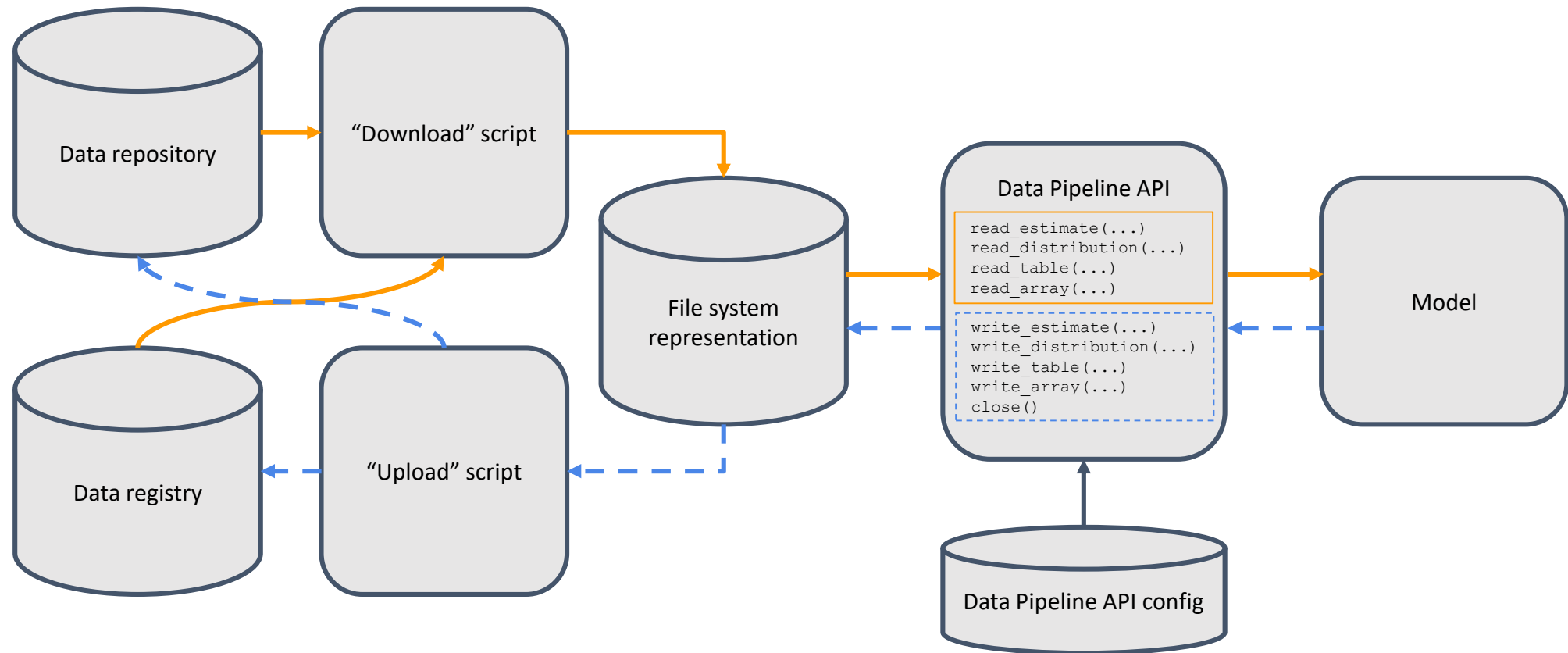
Working on processing additional data products, generating upload templates, and checking robustness of code

Data pipeline for modelling

Richard Blackwell, Chris Hughes, Vino Mano, Rafael de Almeida (Man Group), Ian Hinder (Manchester), Kristian Zarębski (UKAEA), Alex Robson, Sean Lovett (Invenia)



https://github.com/ScottishCovidResponse/data_pipeline_api



Data registry

Jonathan Hollocombe and Andrew Lahiff (UKAEA)



<https://github.com/ScottishCovidResponse/data-registry>

- Available at data.scrc.uk
- Provides web-page view of stored data and information about registry API.
- API can be seen at data.scrc.uk/api
- More info available at data.scrc.uk/docs

SCRC data registration and management system Links ▾ Logged in as: jholloc [logout](#)

[Home](#)

Data Products

Namespace	Name	Version
EERA	fixed-parameters/K	0.1.0
EERA	fixed-parameters/T_hos	0.1.0
EERA	fixed-parameters/T_inf	0.1.0
EERA	fixed-parameters/T_lat	0.1.0
EERA	fixed-parameters/T_rec	0.1.0

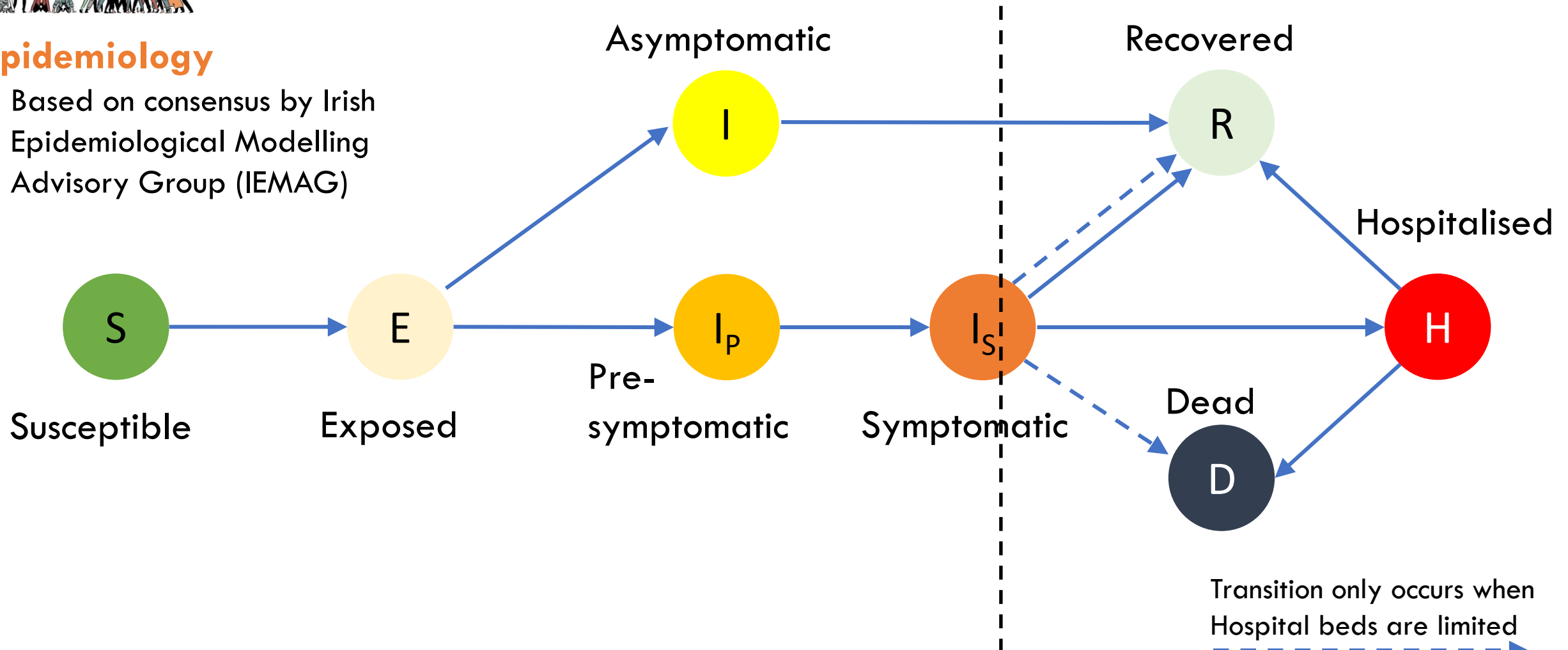
Epidemiology

Louise Matthews (Glasgow / EPIC)



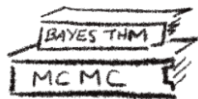
Epidemiology

- Based on consensus by Irish Epidemiological Modelling Advisory Group (IEMAG)



Transition only occurs when Hospital beds are limited

Inference and model assessment



Glenn Marion (BioSS / EPIC)

Inference: quantify epidemic and effect of interventions

- focus on simulation-based methods (model agnostic + parallelisable)
- But also data-augmentation MCMC methods (BEEPmbp)
- And emulation for Uncertainty Quantification and model fitting

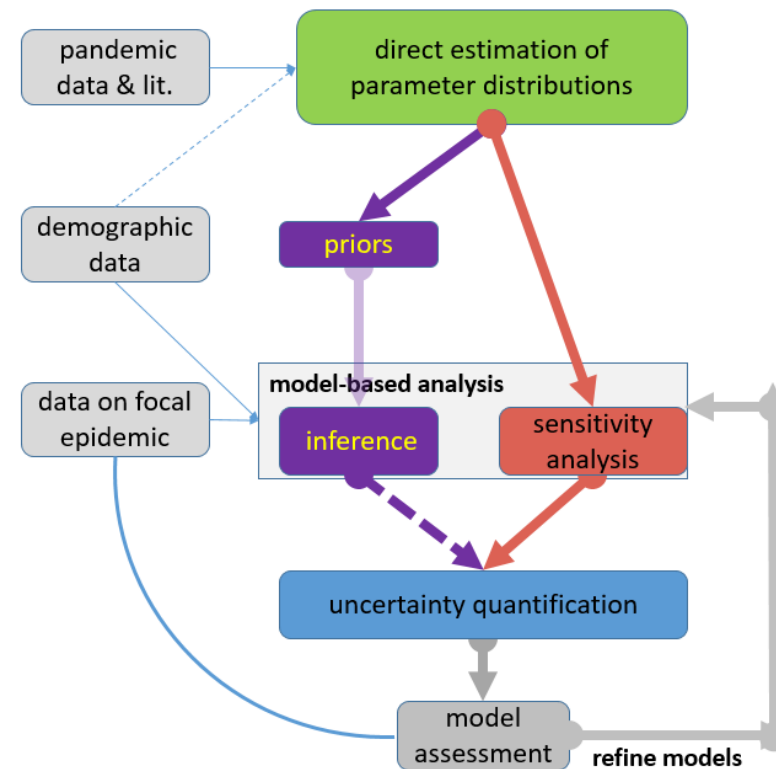
Model assessment: better models

A key issue moving forwards

- statistical fit (evidence, residuals, DIC etc.) and predictions
- model misspecification
- **validation** within and btw model/code-base

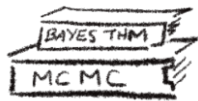
Now in more detail from

- ❑ Ben Swallow (UoG)
- ❑ Rikiya Takahashi and Fedor Gorokhovik (Man Group)



**Inference* + SA > quantify uncertainty
> model assessment > better models**
*focus on Bayesian inference

Inference and model assessment



Ben Swallow (UoG)



Inference: quantify unknown epidemiological parameters

- Again focus on simulation-based methods (easily(?) parallelisable)
- Uses an ABC-SMC algorithm to estimate 6 parameters on 1 km² grid of Scotland
- **quantify uncertainty** in model output and parameters
- Flexible to varying data outputs and thresholds

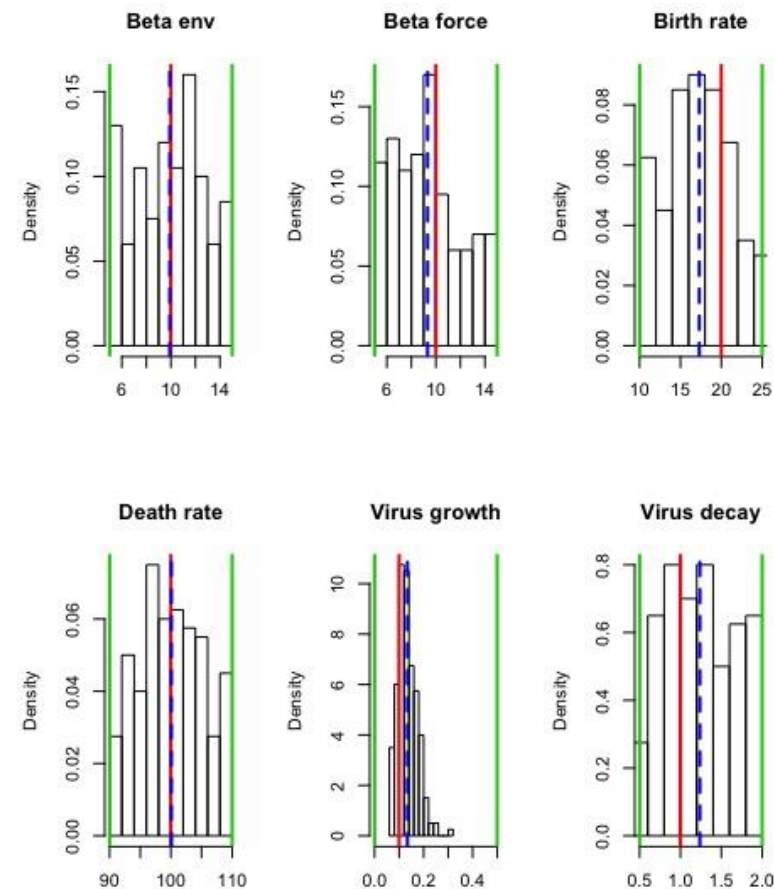
Model testing

- Using data from model simulations at known parameter values we are able to retrieve the true values
- Stochastic model so repeated runs at the same parameter values
- Tested and seems all parameters retrieved using either deaths or symptomatic cases

Issues/open questions

- Link to data pipeline
- Incorporate priors beyond uniform
- Consolidate multiple compartment summaries

Bayesian inference: simulated data



Red solid line = true parameter value
Blue dashed line = posterior mean



Inference and model assessment



Rikiya Takahashi and
Fedor Gorokhovik (Man Group)

Sensitivity Analysis

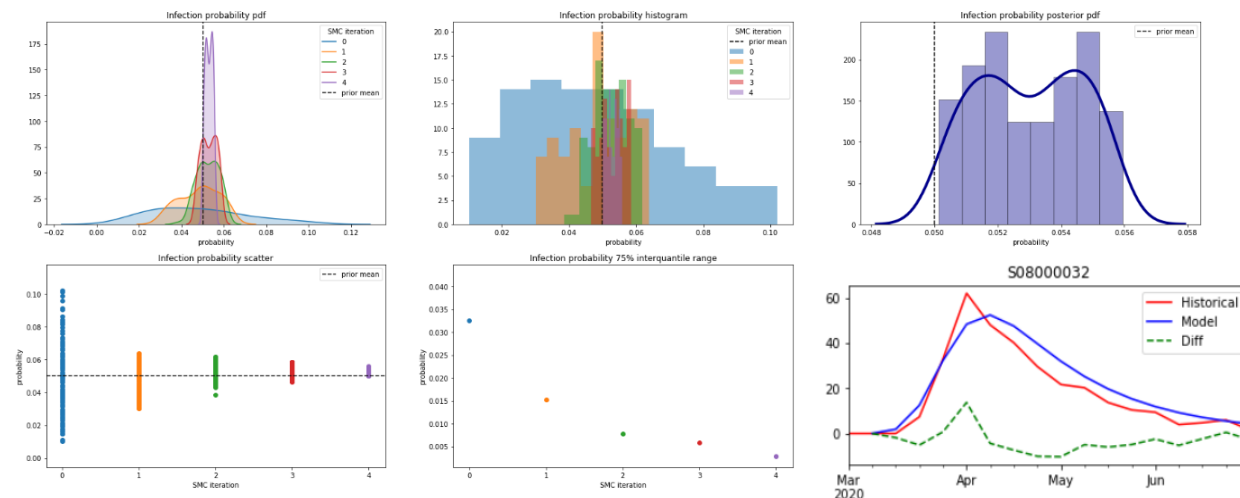
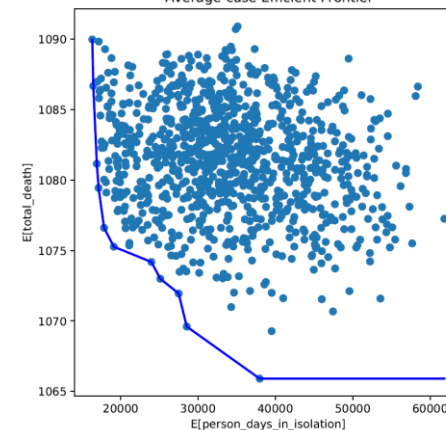
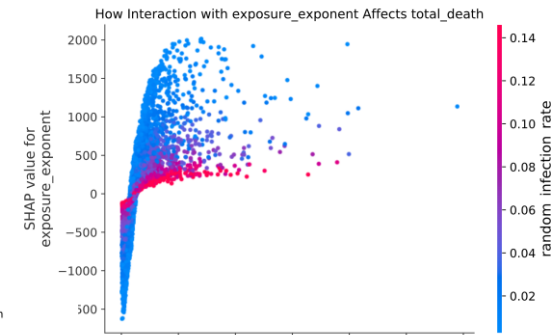
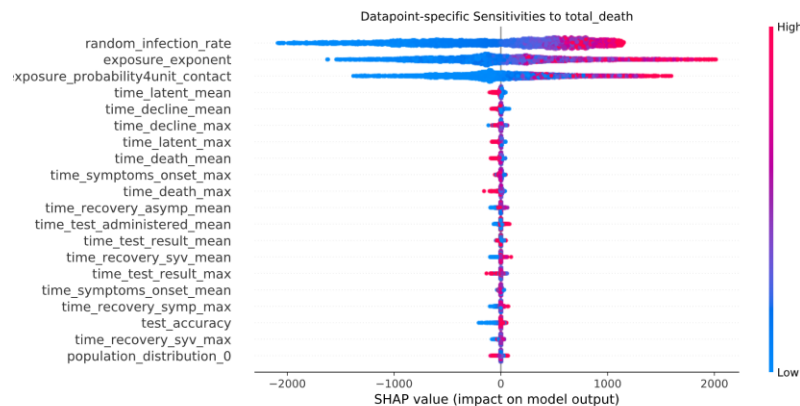
- Reparametrisation of exposure-related disease parameters in Contact Tracing Model
- Non-linear regression and SHAP analysis for total number of death and peak number of severely symptomatic persons
 - SHAP: SHapley Additive exPlanations (Lundberg & Lee, 2017)
- Efficient frontier in optimisation of the policy parameters under uncertainty

Open Issues Real-data-based posterior in policy optimiser, distributional policy parameter recommendations instead of point estimates

Inference

- ABC-SMC used to infer infection probability and initial infections in Simple network sim (contact multipliers to come)
- To be used in conjunction to stochastic mode to have more realistic posterior std dev
- Fitted on historical death per healthboard in Scotland
- Other parameters sources from external sources to be as credible as possible
- Wide priors chosen from Beta, Lognormal distribution

Open Issues Improved rejection mechanism, parallelization of inference



VIS

27 VIS/VA volunteers, plus Min Chen (Oxford U.)



<https://github.com/ScottishCovidResponse/rampvis-ui>

[rampvis-api](#)

[srcv-vis-analytical](#)

[srcv-vis-modelling](#)

[Dr. Alfie Abdul-Rahman](#), KCL

Dr. Daniel Archambault, Swansea

Dr. Benjamin Bach, Edinburgh

Dr. Rita Borgo, KCL

[Professor Min Chen](#), Oxford

Professor Jason Dykes, City

Dr. Hui Fang, Loughborough

Elif Firat, Nottingham

Dr. Radu Jianu, City

Professor Nigel W. John, Chester

Dr. Saiful Khan, Horus Security

Dr. Robert S Laramée, Nottingham

Dr. Stella Mazeri, Edinburgh

Benjamin Nash

Dr. Phong H. Nguyen, Red Sift Ltd.

Dr. Helen C. Purchase, Glasgow

Dr. Dylan Rees, Swansea

Dr. Panagiotis Ritsos, Bangor

Professor Jonathan Roberts

James Scott-Brown, Edinburgh

Dr. Aidan Slingsby, City

Dr. William Teahan, Bangor

Dr. Thomas Torsney-Weir, Swansea

Dr. Cagatay Turkay, Warwick

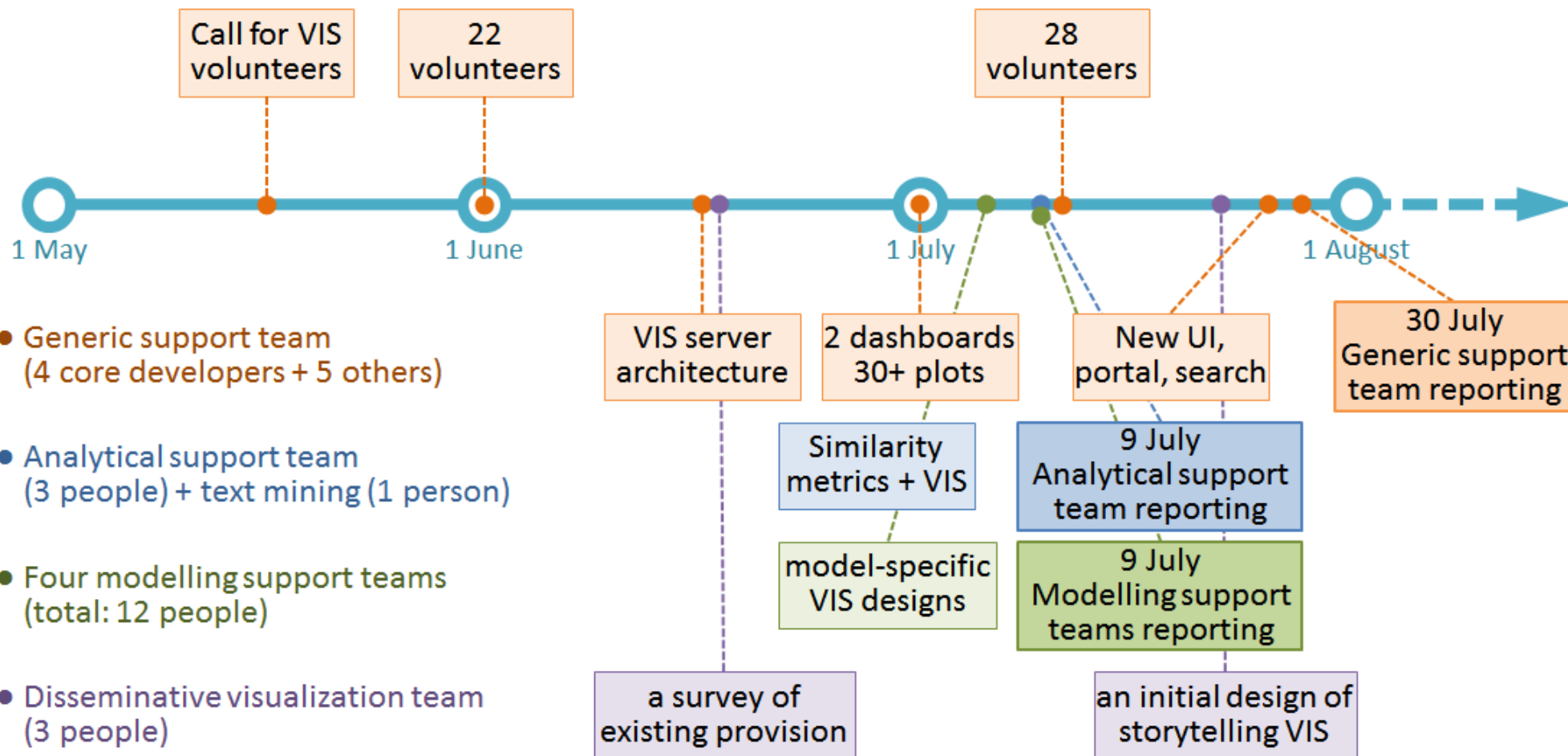
Dr. Franck Vidal, Bangor

Qiru Wang, Nottingham

Tianci Wen, Bangor

Professor Jo Wood, City

Dr. Kai Xu, Middlesex



VIS

27 VIS/VA volunteers, plus Min Chen (Oxford U.)



<https://github.com/ScottishCovidResponse/rampvis-ui>

[rampvis-api](#)

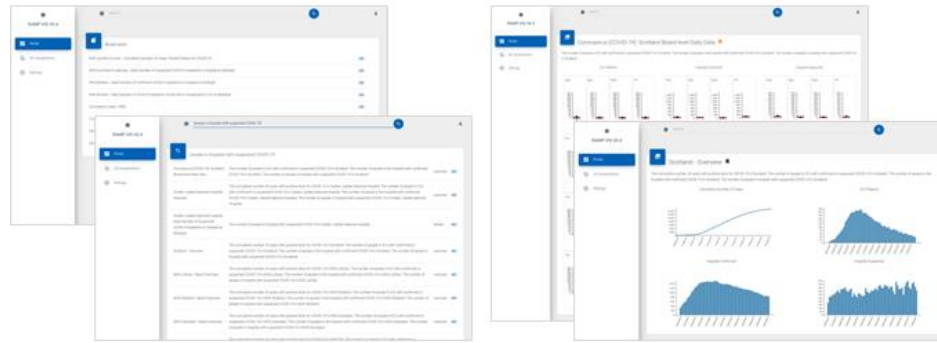
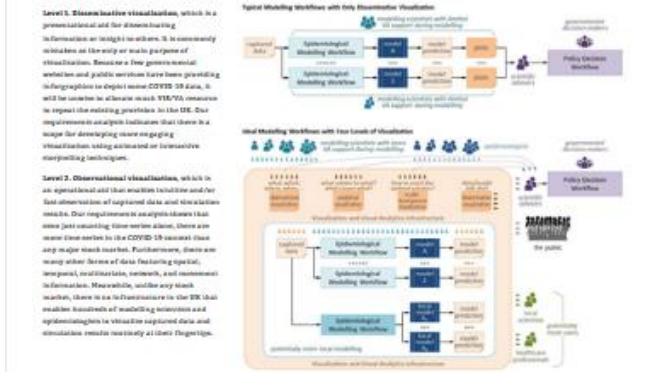
[scrc-vis-analytical](#)
[scrc-vis-modelling](#)



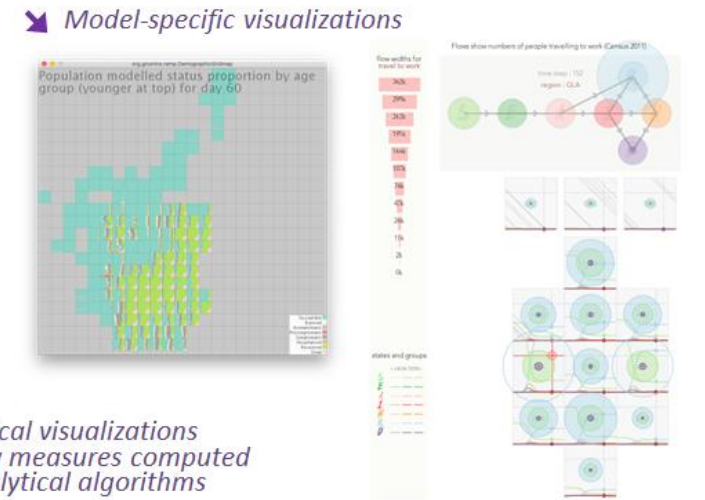
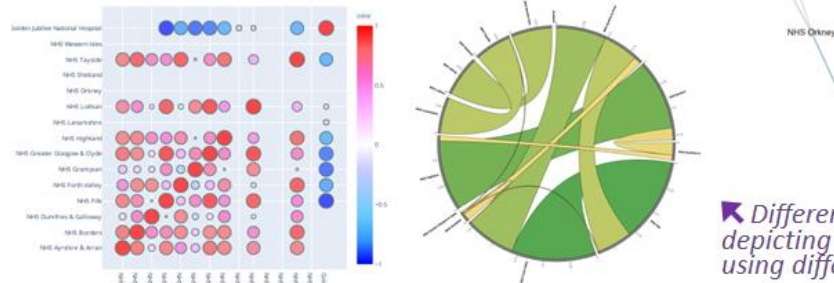
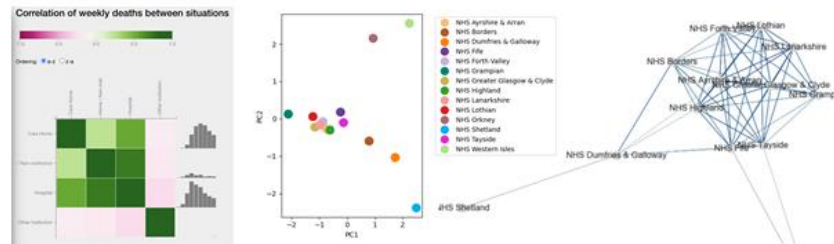
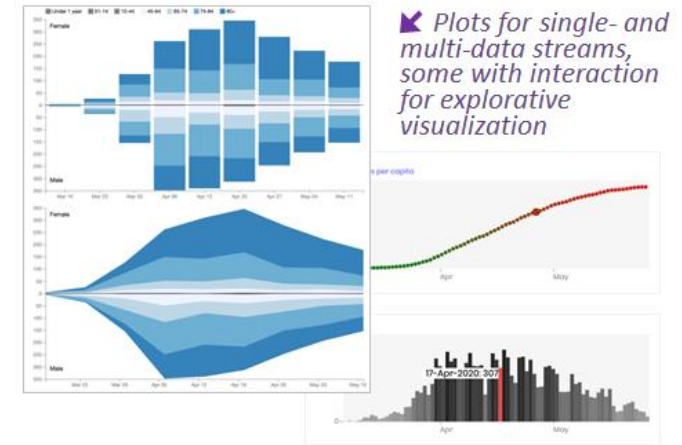
The above image is by Henning Heideberg from Statista.

RAMP VIS is a group of [volunteers](#) specialized in Data Visualization and Visual Analytics, who answered a call to support the working scientists and epidemiologists in the [British COVID-19 Response Consortium \(BCRC\)](#).

Most people struggle in the dynamic task domain and research use faces in driving. Many were hit by a substantial amount of 7 years ago evidence and analysis, with many are still being improved today. The research can usually be distributed in a collection of modelling scenarios in data collection, observation, and analysis (hypothesis formulation and model development, validation, deployment, monitoring, and improvement). In analyzing COVID-19, such a collective effort in research has to be completed in a period of weeks and months. Without adequate visualization (VE) and visual analytics (VA) support, an existing workflow typically requires modelling scenarios and epidemiologists to spend more time analyzing the results. Some data, to observe some data and make the frequent observations, is only made via data mining algorithms (usually without keeping track of potential limitations and errors, or is only made via tables without using algorithms to explore different options). Hence, there is an urgent need to increase the VIS/VA capacity to track rapid model development workflows, and to maintain the precision of data capacity throughout the period of analyzing the COVID-19 pandemic. There are many different visualization tasks in different model development workflows. They can be broadly categorized into four levels according to the size and complexity of the model space for an optimal device.



The RAMP VIS interface with personal portals, a search facility, dashboards.



Different analytical visualizations depicting similarity measures computed using different analytical algorithms

Open Epidemiology



<https://github.com/ScottishCovidResponse>

Responsibility

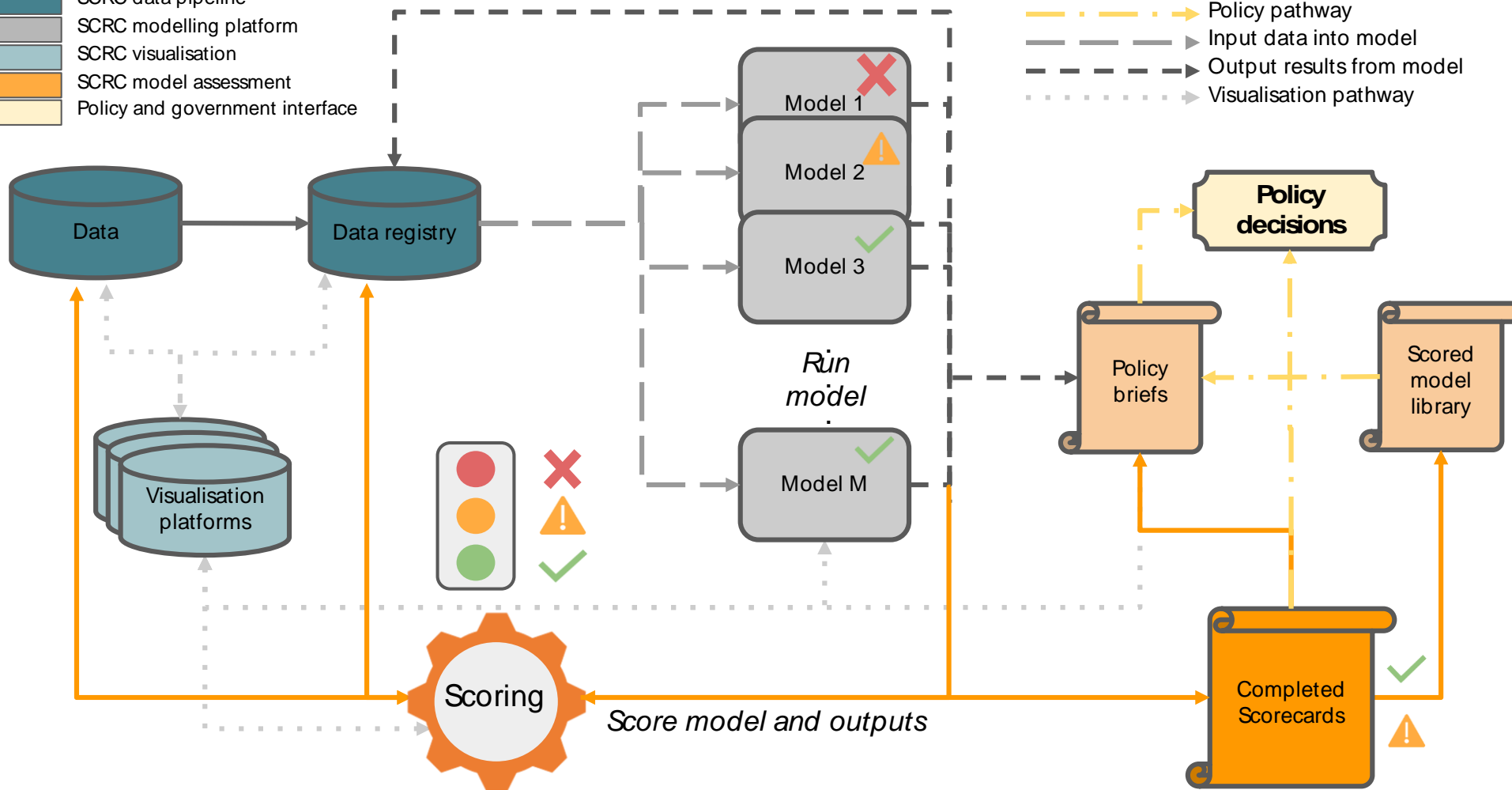
	SCRC data pipeline
	SCRC modelling platform
	SCRC visualisation
	SCRC model assessment
	Policy and government interface

Action

	Modelling platform assessment
	Policy pathway
	Input data into model
	Output results from model
	Visualisation pathway

Requirements

- Reproducible
- Traceable
- Validated
- Open source
- Open data



Administration

Lynn Esson (Glasgow) and Yasmin Abdalla (Edinburgh)

Rachel O'Brien (Glasgow)

Mark Meenan (Glasgow)



Requirements

- Extreme patience



<https://github.com/ScottishCovidResponse>

Universities of Glasgow, Edinburgh, SRUC, Sheffield, Manchester,
St Andrews, Stirling, Strathclyde, Heriot-Watt, Southampton,
Cambridge, Oxford, Swansea, Bangor, Nottingham,
Middlesex, City, U. of London, Warwick, Chester,
King's College London, Loughborough, EPIC,
Biomathematics and Statistics Scotland,
UK Atomic Energy Authority, Met Office,
Software Sustainability Institute,
Man Group, Invenia,
Horus Security,
Red Sift Ltd,
and others

