

Modelling the impact of noise on wastewater-based environmental surveillance

Jeremy Bingham

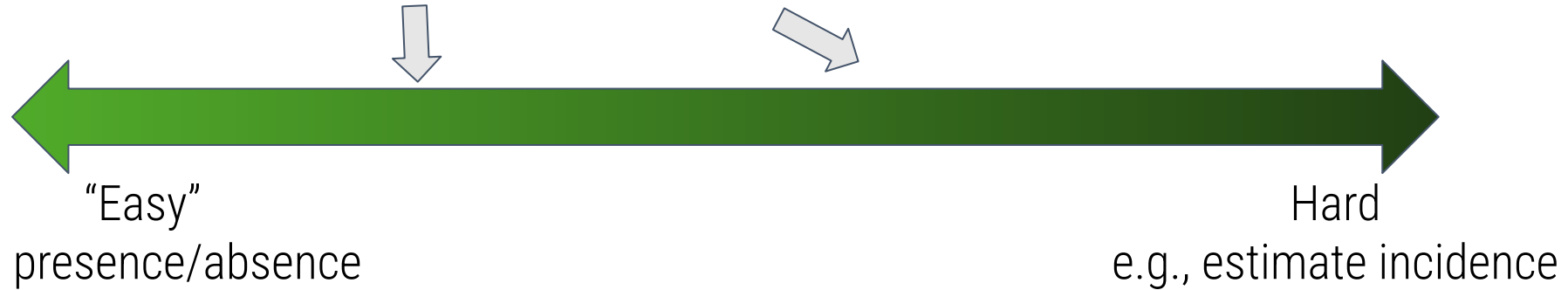
2024-06-26



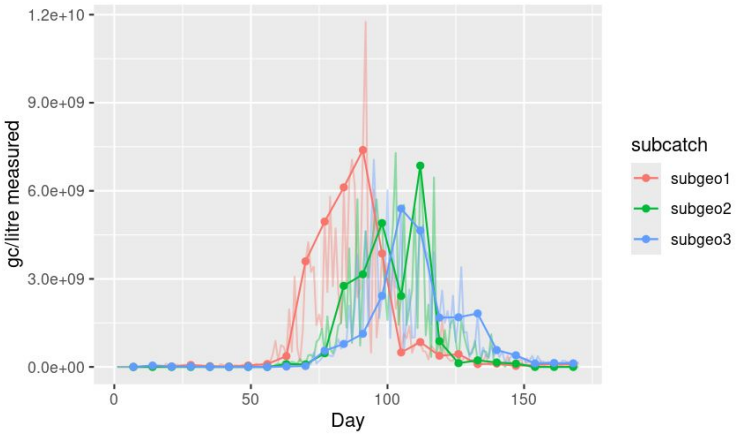
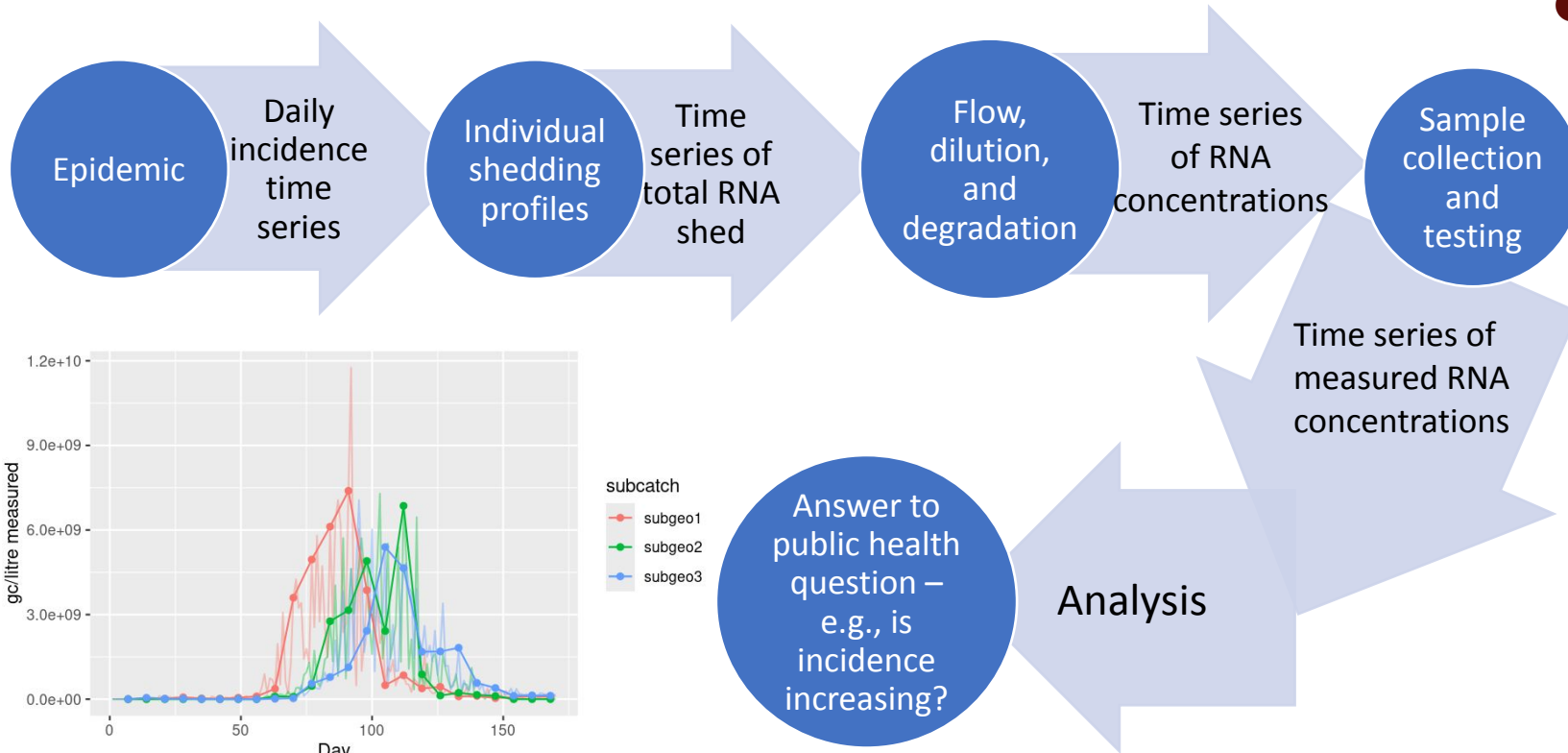
Motivation



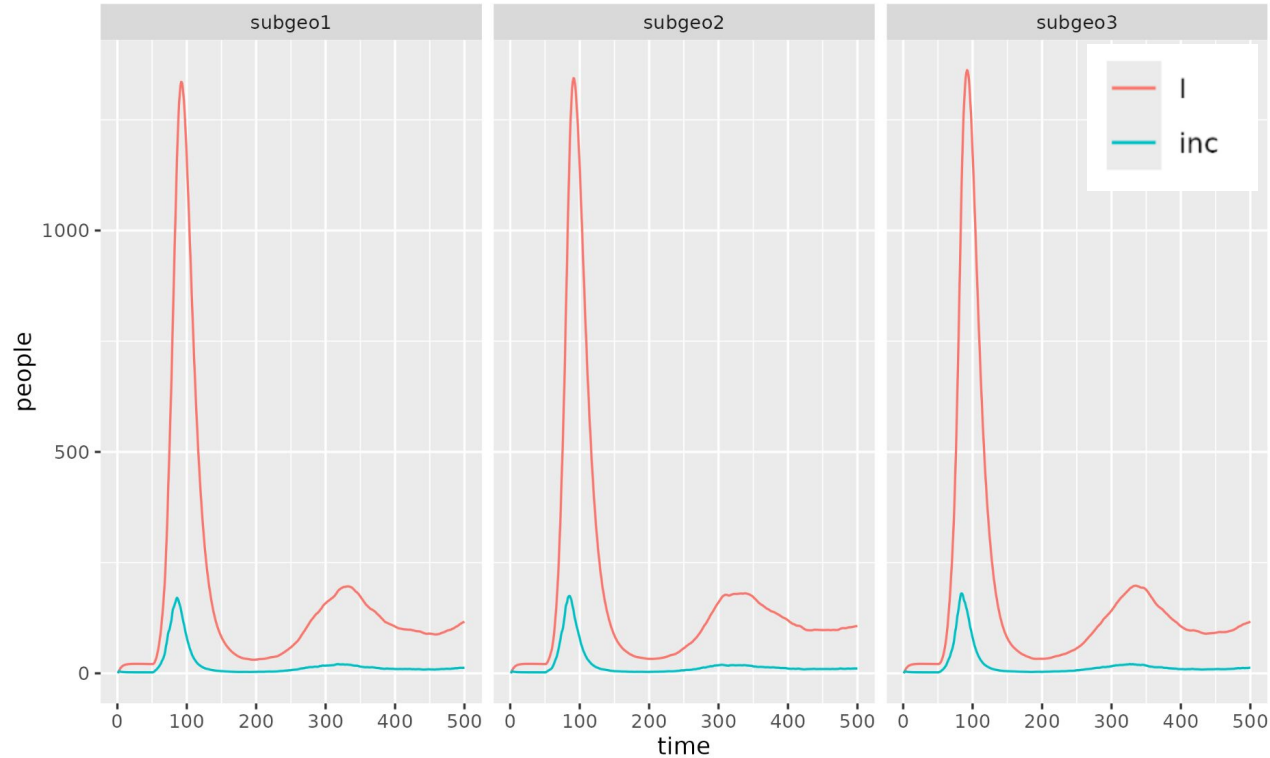
- Wastewater contains valuable information
- Mixed *practical* success
- Misalignment of reality vs expectations



Modelling pipeline: steps



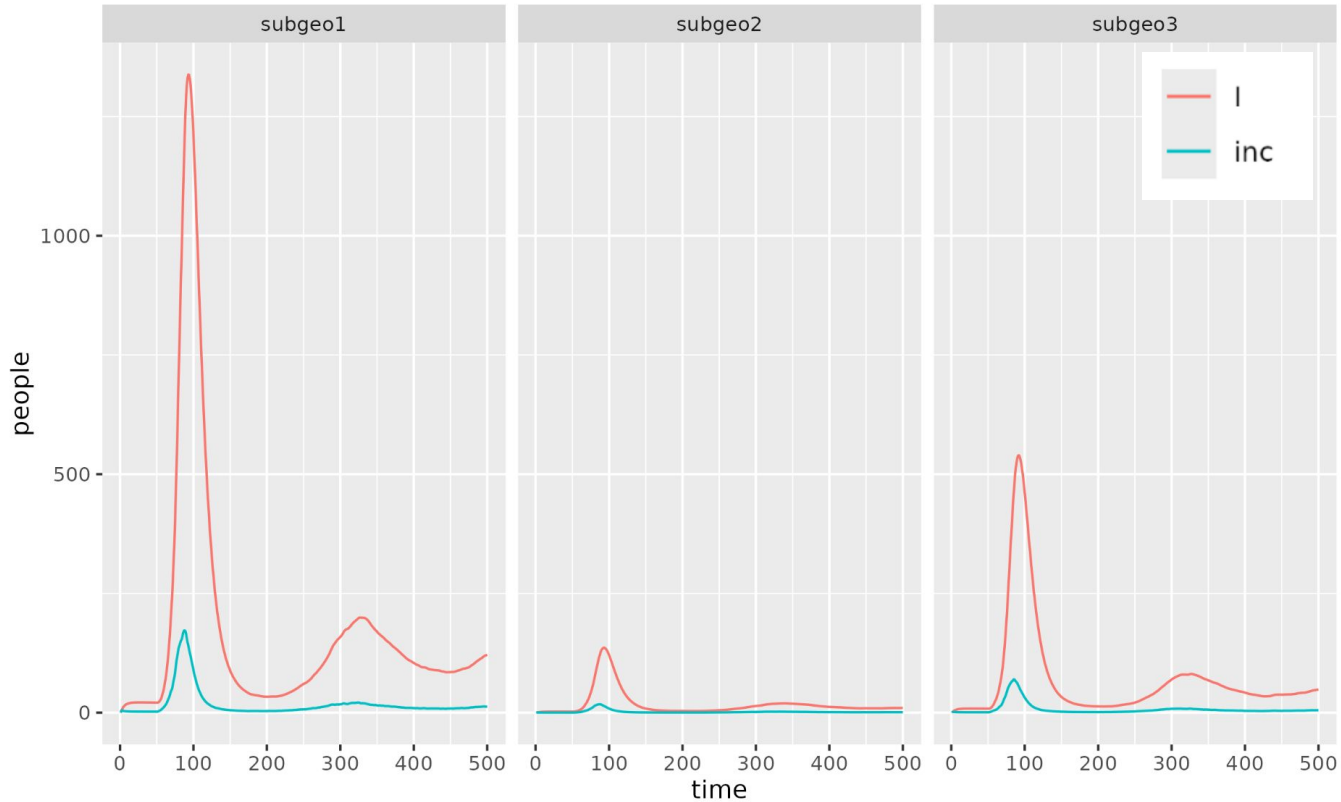
Base case: 3 nearly-identical epidemics



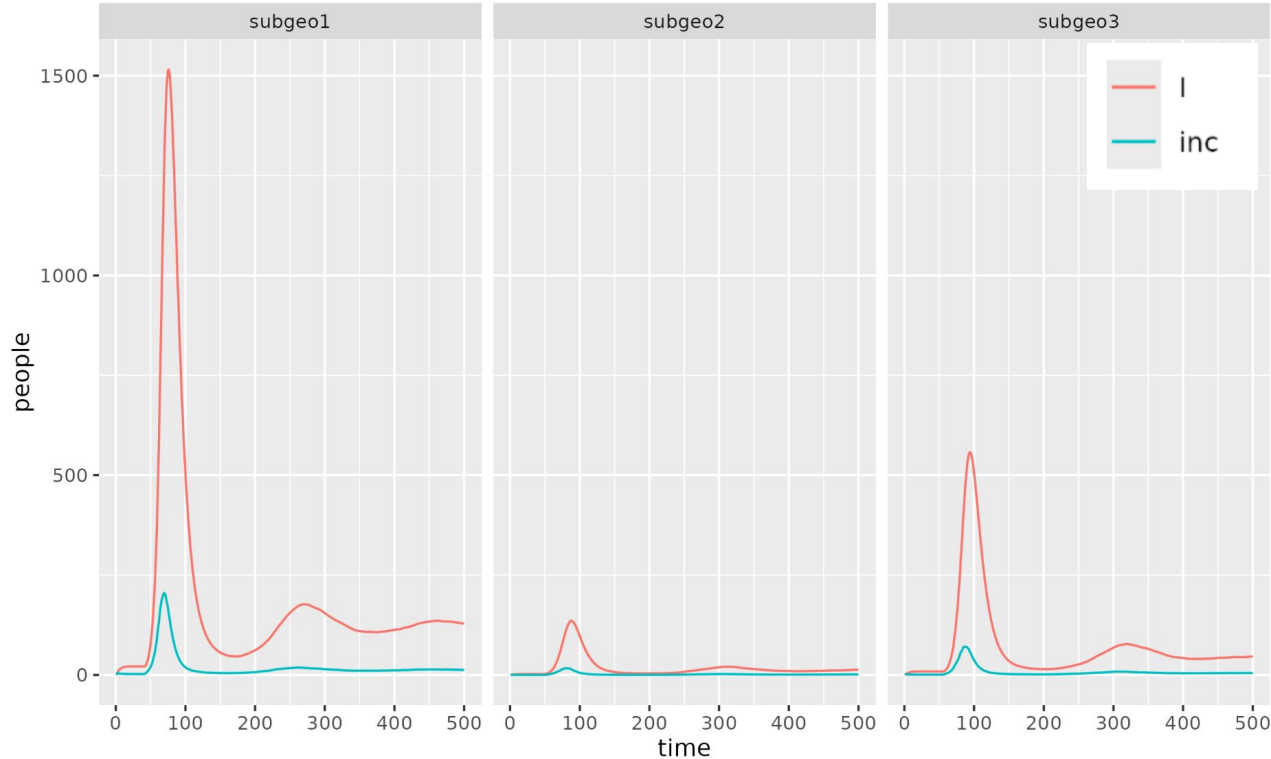
SEIRS model

Initial lull (low transmission) followed by higher transmission

Second case: same epidemic parameters, different population sizes (5000, 500, 2000)



Third case: different population sizes, start times, and transmission rates



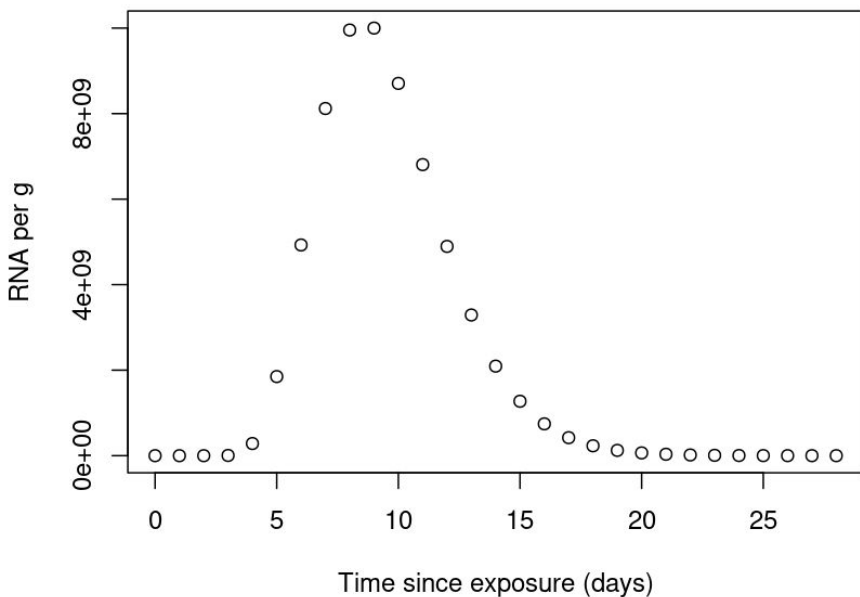
Epidemics “start” 1 & 2 weeks apart

- Large population, high transmission
- Small population, low transmission
- Medium population, moderate transmission

Shedding

Individual-level shedding

Gamma distributed concentration over time **RNA/g**



Flow/Dilution



White noise around a mean **per-person dilution**

Input: Total RNA shed (daily # gene copies)

Output: gene copies/litre

Sample collection and measurement

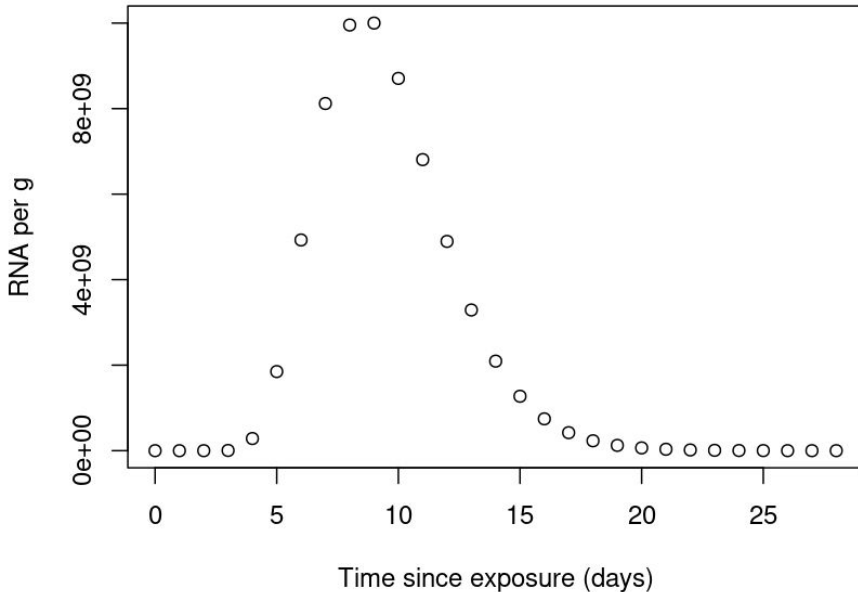
White noise around daily mean

Observation daily/4-days/weekly

Shedding

Individual-level shedding

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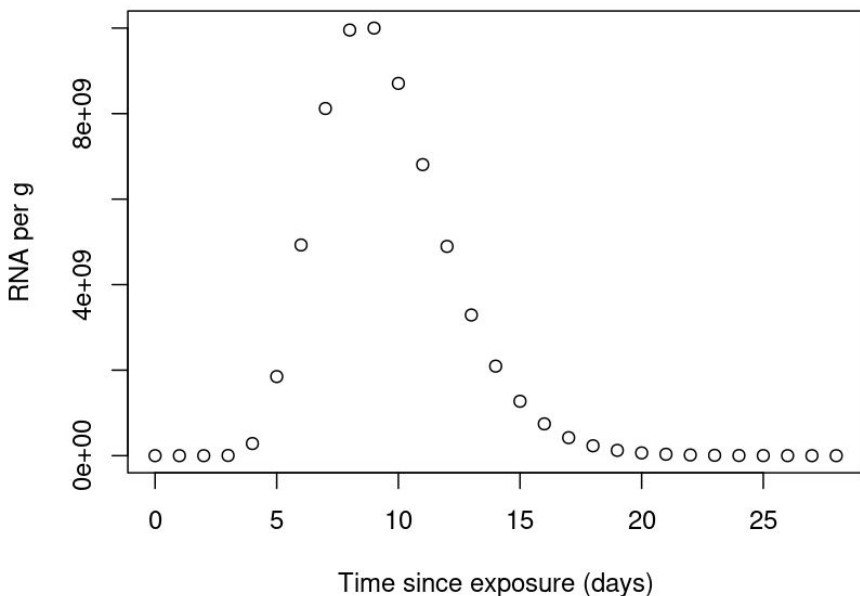
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Analysis



How often do we “detect” a difference between two catchment populations using a paired t-test,

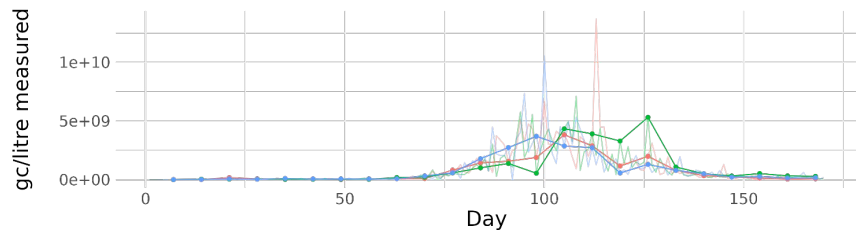
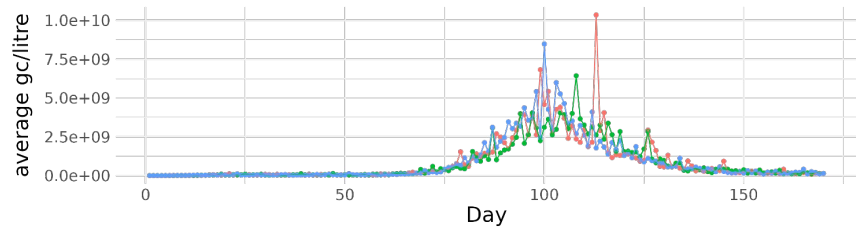
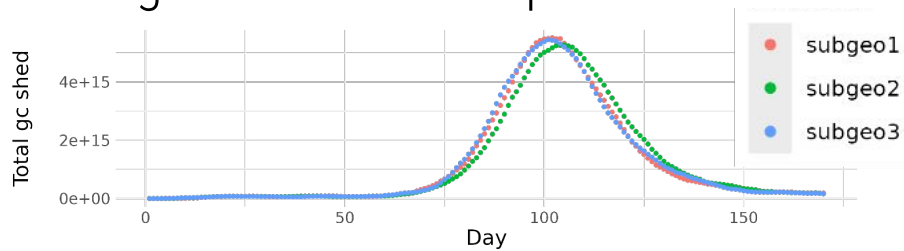
- when epidemic dynamics are identical?
- when epidemic dynamics are different?

Are the different sub-catchments likely to be reliably differentiable for answering public-health-relevant questions?

Example: three identical epidemics

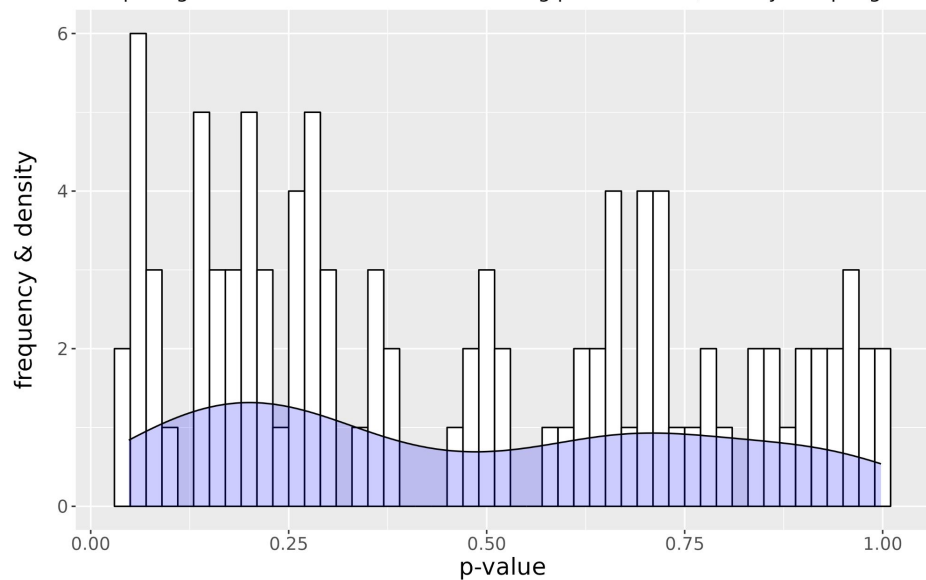


“High” noise in all steps



p-values roughly evenly spread
(and approx 5% under 5%)

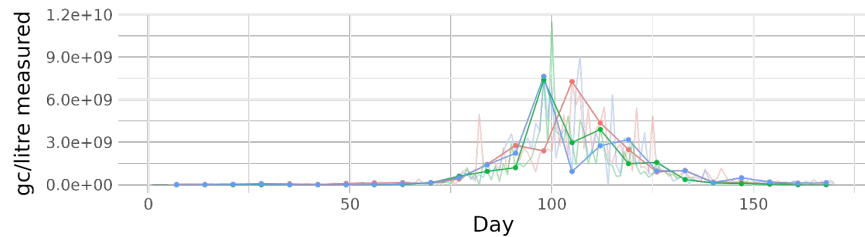
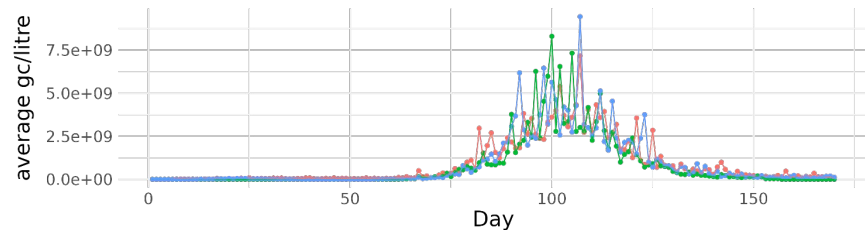
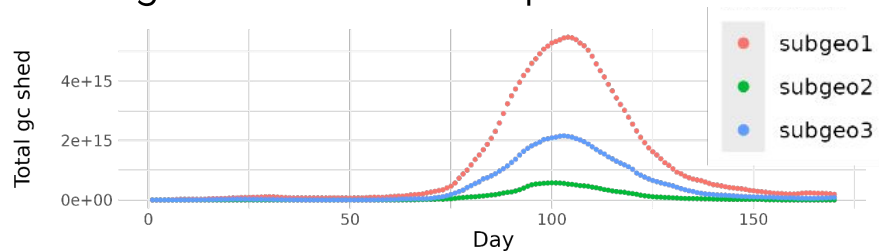
Comparing subcatchments one and two using paired t-tests, weekly sampling



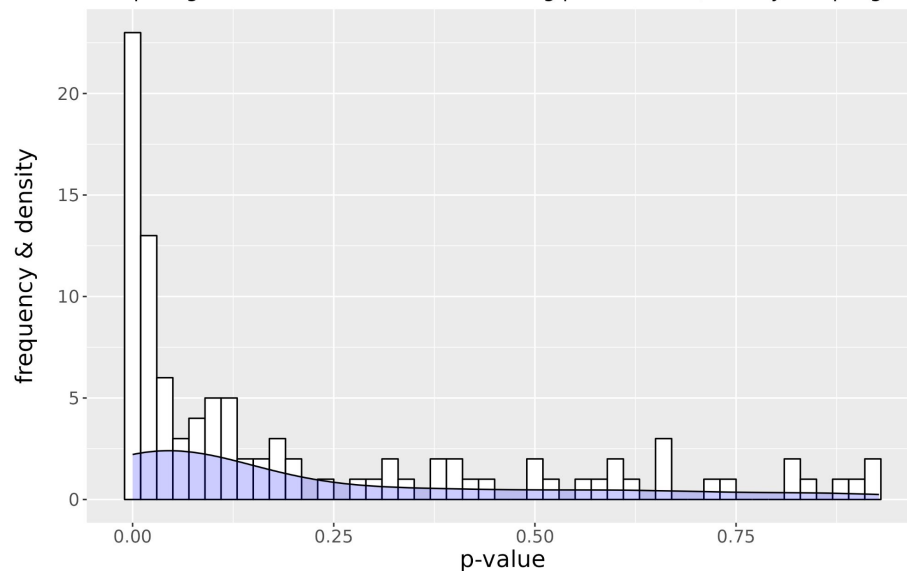
Example: three similar epidemics (Δ pop sizes)



“High” noise in all steps



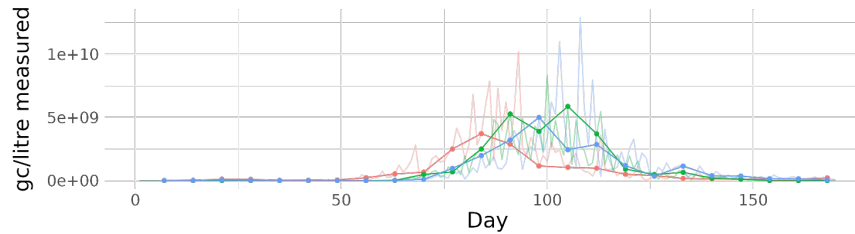
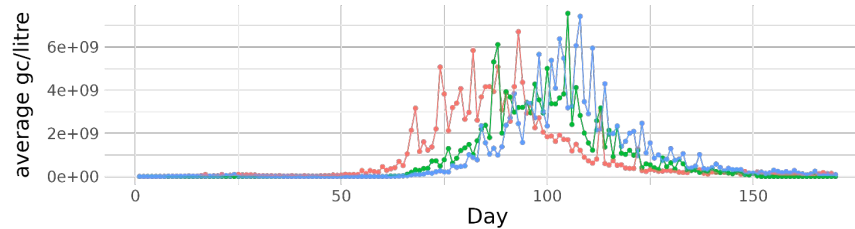
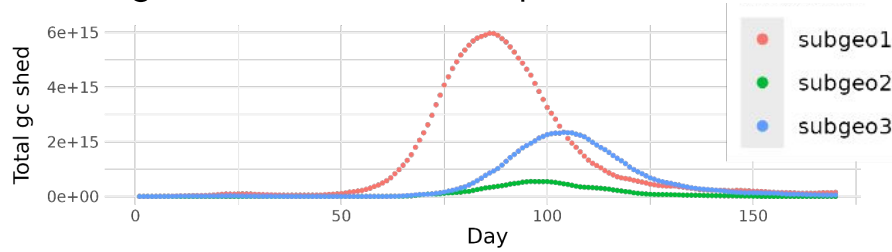
Comparing subcatchments one and two using paired t-tests, weekly sampling



Example: three different epidemics

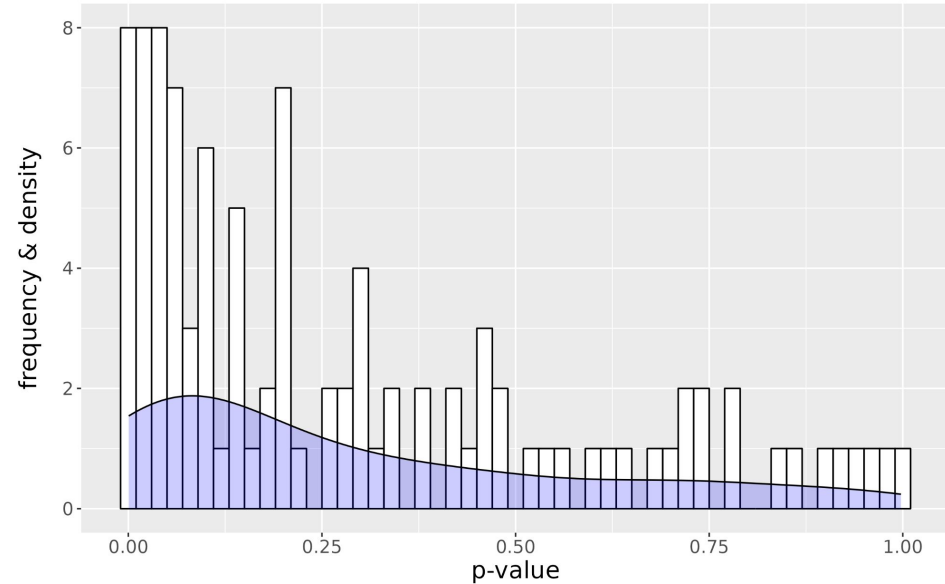


“High” noise in all steps



Too much noise to reliably detect differences, even over long timespan

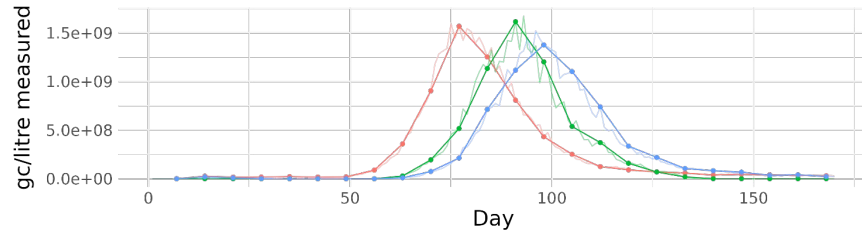
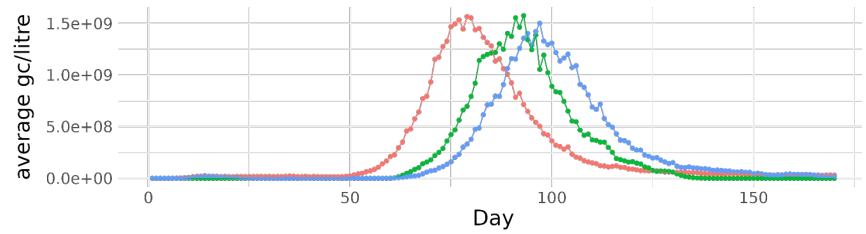
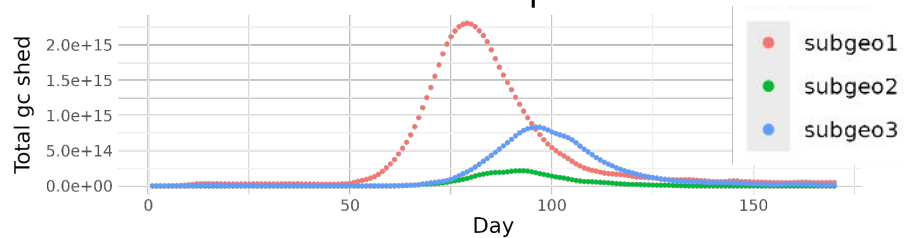
Comparing subcatchments one and two using paired t-tests, weekly sampling



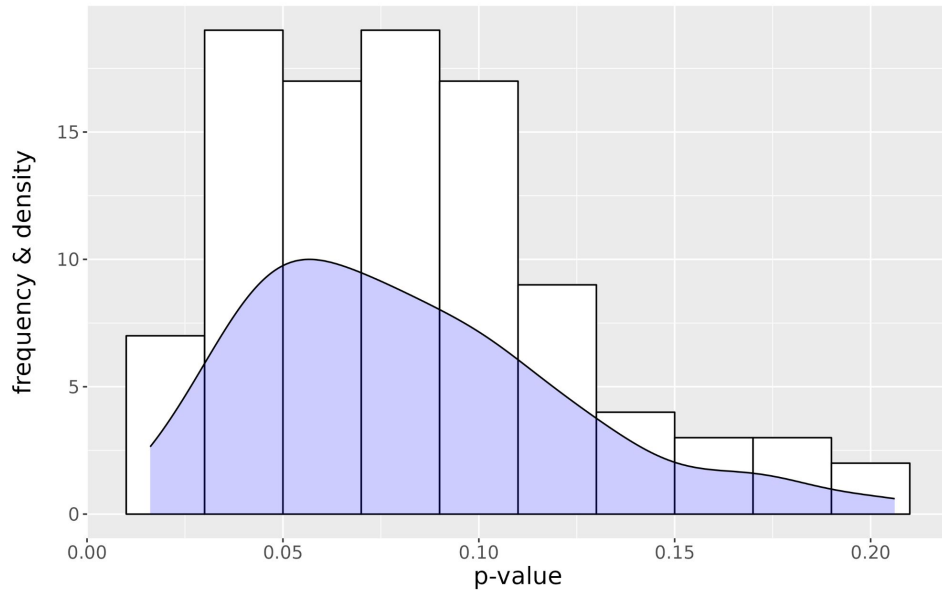
Example: three different epidemics



“Low” noise in all steps



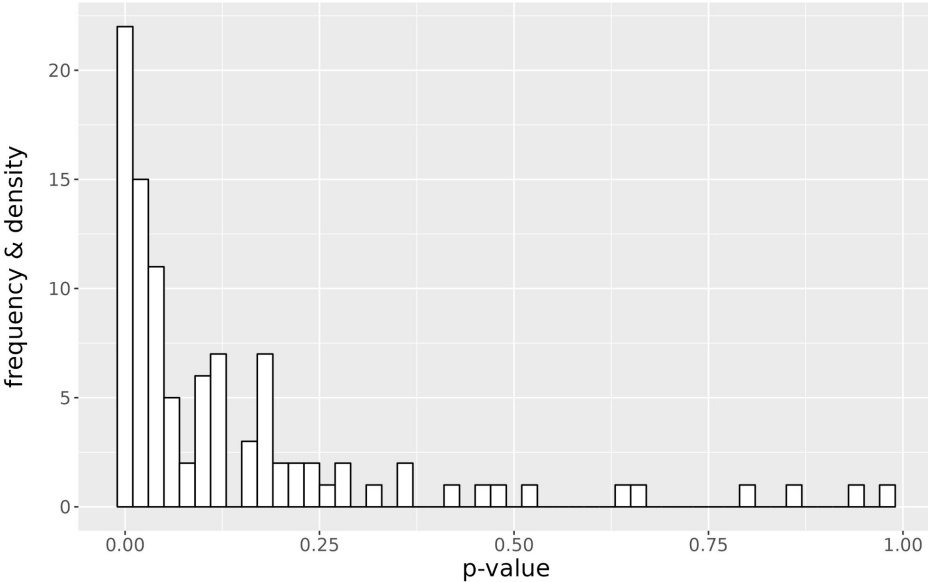
Comparing subcatchments one and two using paired t-tests, weekly sampling



What if we can't reduce the noise?

Increase sampling frequency
from daily to every four days:

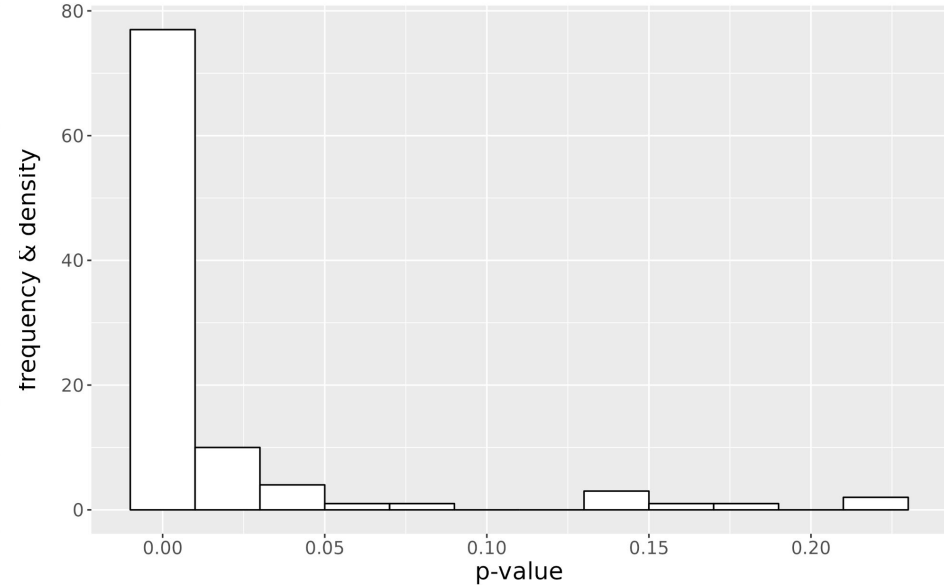
Comparing subcatchments one and two using paired t-tests, fourdays sampling



With daily sampling we can
reliably tell catchments
apart!



Comparing subcatchments one and two using paired t-tests, daily sampling



Key takeaways

- Characterise noisiness of system as early as possible
- Set clear expectations
- Design surveillance for specific questions

Next steps

- Incorporate data on South African systems and catchment populations
- Test practical public health question
 - e.g. “is a new wave starting”
- Leverage nested catchments to filter outliers

Special thanks to



- Kathleen O'Reilly (LSHTM)
- Cari van Schalkwyk (SACEMA)
- Kerrigan McCarthy (NICD)
- Gauteng City Regional Observatory
- Bill and Melinda Gates Foundation

and thank *you* for listening!

Shedding

Individual-level shedding:

Gamma distributed concentration over time **RNA/g**

Start shedding at 3 days

Shedding for 25 days

Noise via:

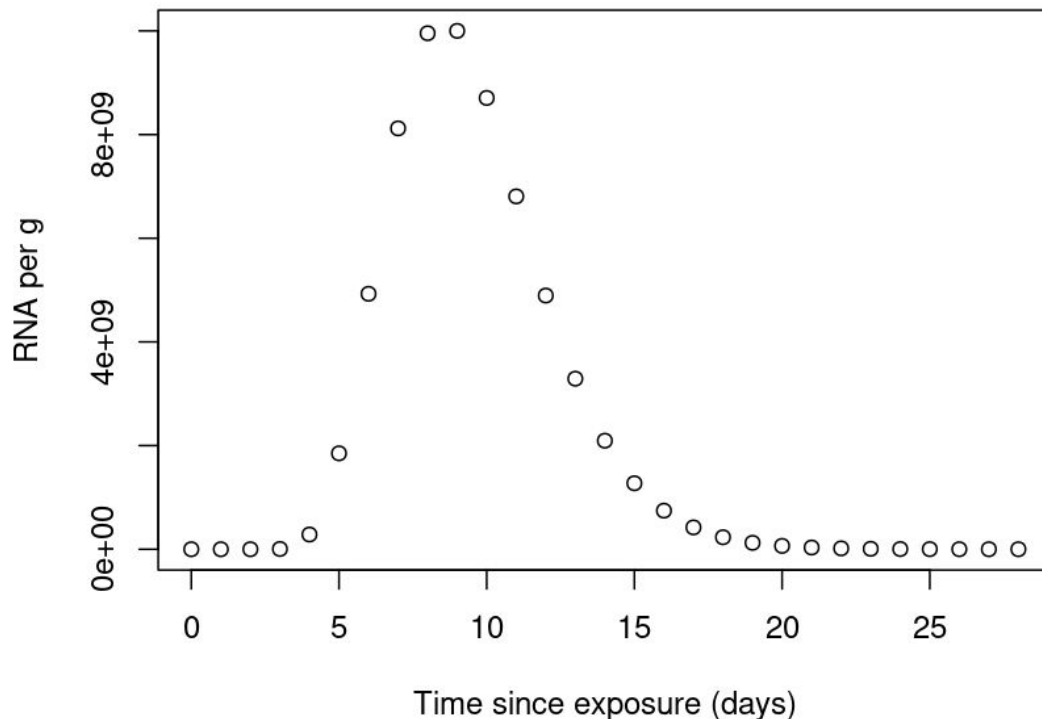
Peak RNA/g (mean 10^{10})

Time to peak shedding (10 days)

“Scale” parameter of gamma distribution (mean 1)

Assume daily **200g** per person

Example of individual shedding profile



Peak RNA/g = 10^{10} ; peak at 10 days;
Start at 3 days; duration of 25 days

Flow/Dilution



White noise around a mean **per-person dilution**

Input: Total RNA shed (daily # gene copies)

Output: gene copies/litre

(**300l per person** per day, slightly above national average of 237l per person per day)

Still open question: how best to measure dilution and characterise the noisiness of dilution estimates?

“Low” noise: Standard deviation = 5% of mean

“Medium” noise: Standard deviation = 10% of mean

“High” noise: Standard deviation = 30% of mean

Sample collection and measurement



Collection and laboratory procedure currently modelled in a single step:

Input: Daily average concentration (gene copies/litre)

Output: Daily/four-days/weekly measured concentration (gene copies/litre)

White noise around **daily mean**

Parameterised via standard deviation as proportion of daily mean

Low noise: 5% of mean

Medium noise: 10% of mean

High noise: 40% of mean