

# The Clue is in the Poo...

*utilising wastewater data for infectious diseases*

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HYGIENE  
& TROPICAL  
MEDICINE



- What happens?
- Examples of using environmental surveillance (ES) data\* for infectious diseases
- Questions to ask of ES data...
- Example of COVID-19
- From research to routine use

*\* Also called, sewage data, wastewater data (depends on context and research specialism)*

# How?

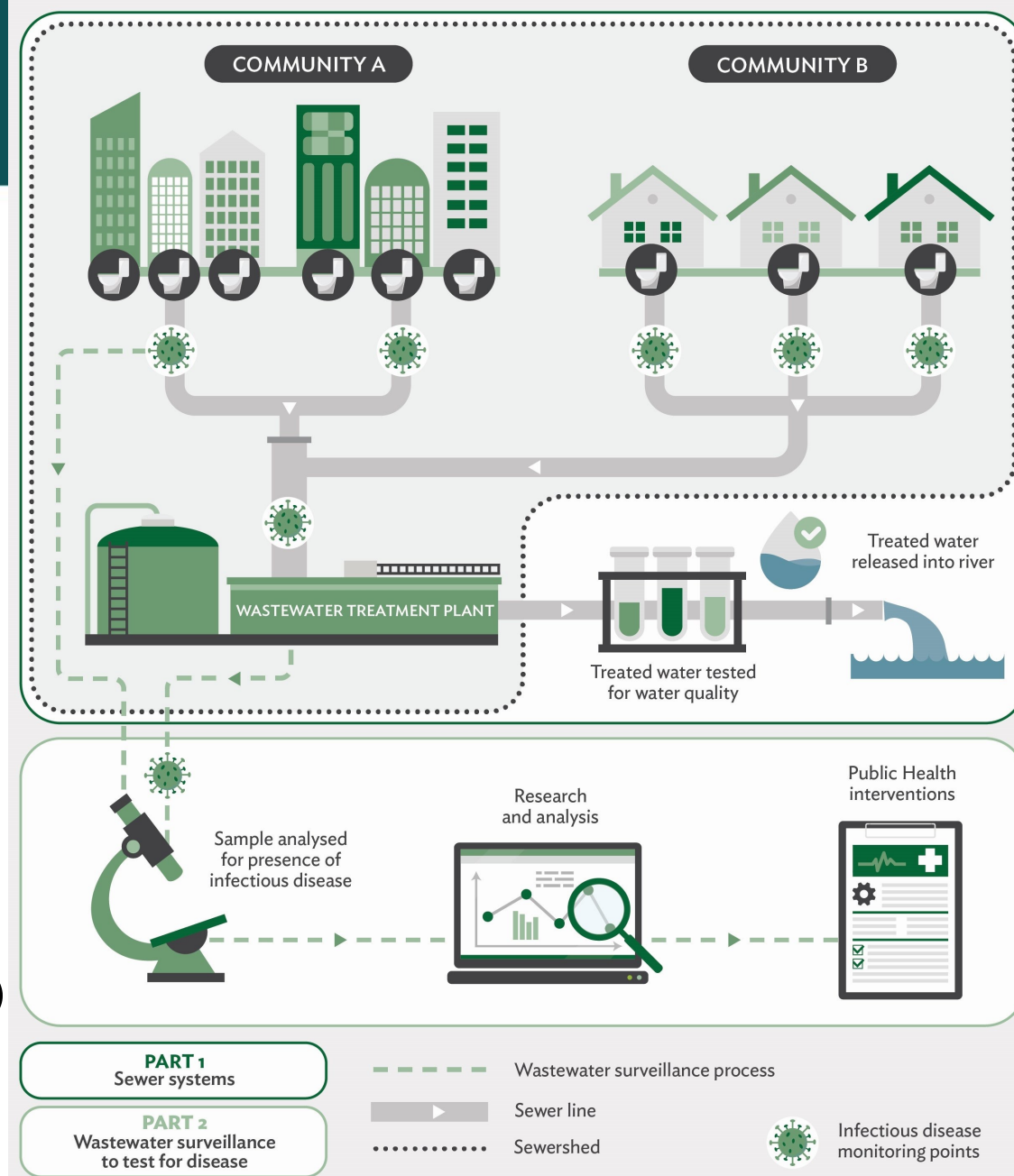
- Many pathogens are shed in feces and urine
  - Part of the transmission cycle (fecal-oral)
  - And RNA/DNA, body removing waste
- Polio eradication
  - Shed live virus (fecal-oral transmission)
  - Virus culture & RNA detection
  - Detection in London sewage in 2022\*
- Typhoid
  - Water & food contamination
- Others
  - COVID-19 (SARS-CoV-2), 'flu, norovirus, measles, cholera, mpox, ...



\* <https://tinyurl.com/wbepolio>

# What happens?

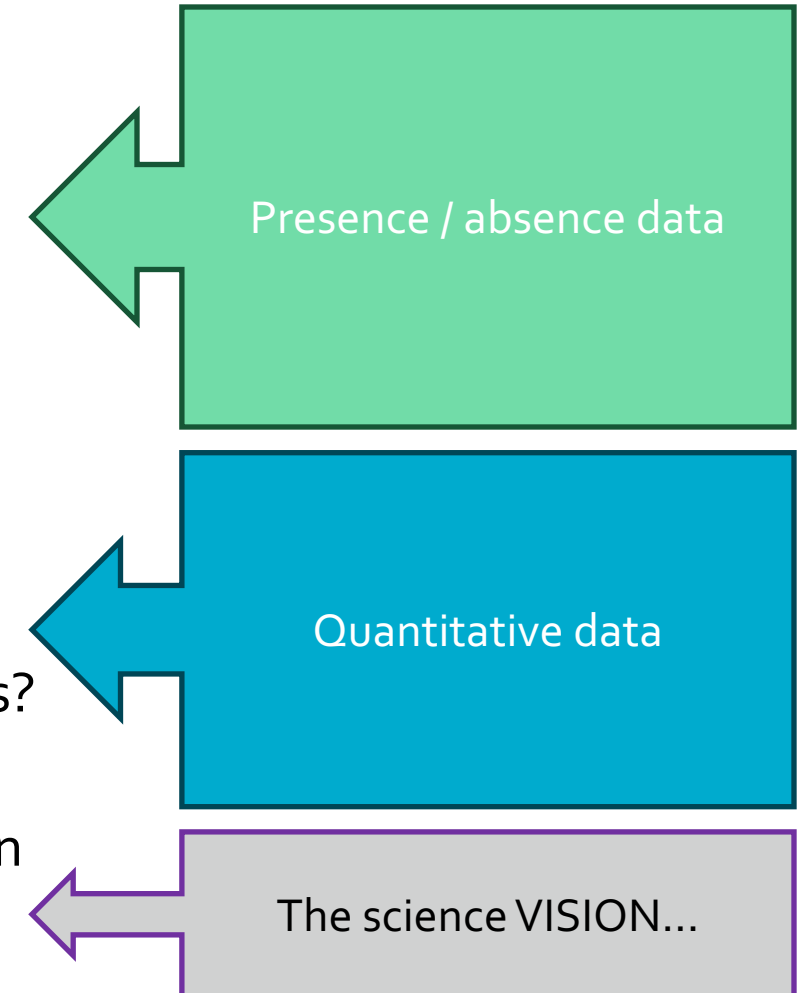
- Select a site for regular sampling
- 'Grab' a sample, or use composite sampler
  - Maybe record site characteristics (eg. flow)
- Take sample to the lab
- Carry out lab analysis:
  - 1) Record 'meta-data' (eg. ammonia)
  - 2) Concentrate and/or extract the sample
  - 3) Identify what you are looking for
    - PCR based analysis (primers...)
    - Virus isolation (cell culture)
    - WGS (eg. nanopore / illumina sequencing)
    - Meta-genomics
- Make sense of the data collected



More details: <http://tinyurl.com/wbegcro>

# Questions to ask of ES data

1. Is pathogen A present?
  - Single or multiple sites
  - For a defined duration
2. Is pathogen A absent?
  - Single/multiple sites, and over a defined duration
3. How has the prevalence of pathogen A changed in time? (1 site)
4. How has incidence of pathogen A changed in time? (1 site)
5. How does prevalence/incidence compare between locations? (multiple sites)
6. What is the incidence of infection (or disease) with pathogen A, estimated by combining clinical and ES data together



# Change in pathogen prevalence

Using a simple model,

**Morvan et al (2022)**

- Prevalence estimation
- Sensitivity analysis

$$P = \frac{C \times Q_p}{S \times V}$$

Concentration (GC/L)  $C$

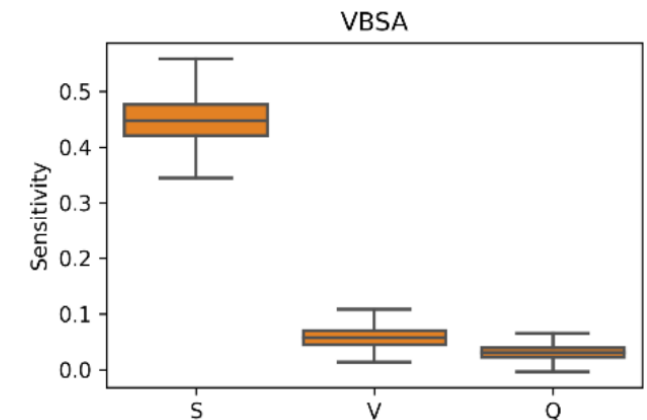
Quantity wastewater (L/day/person)  $Q_p$

Shedding rate (GC/ml)  $S$

Volume of stool (ml/day)  $V$

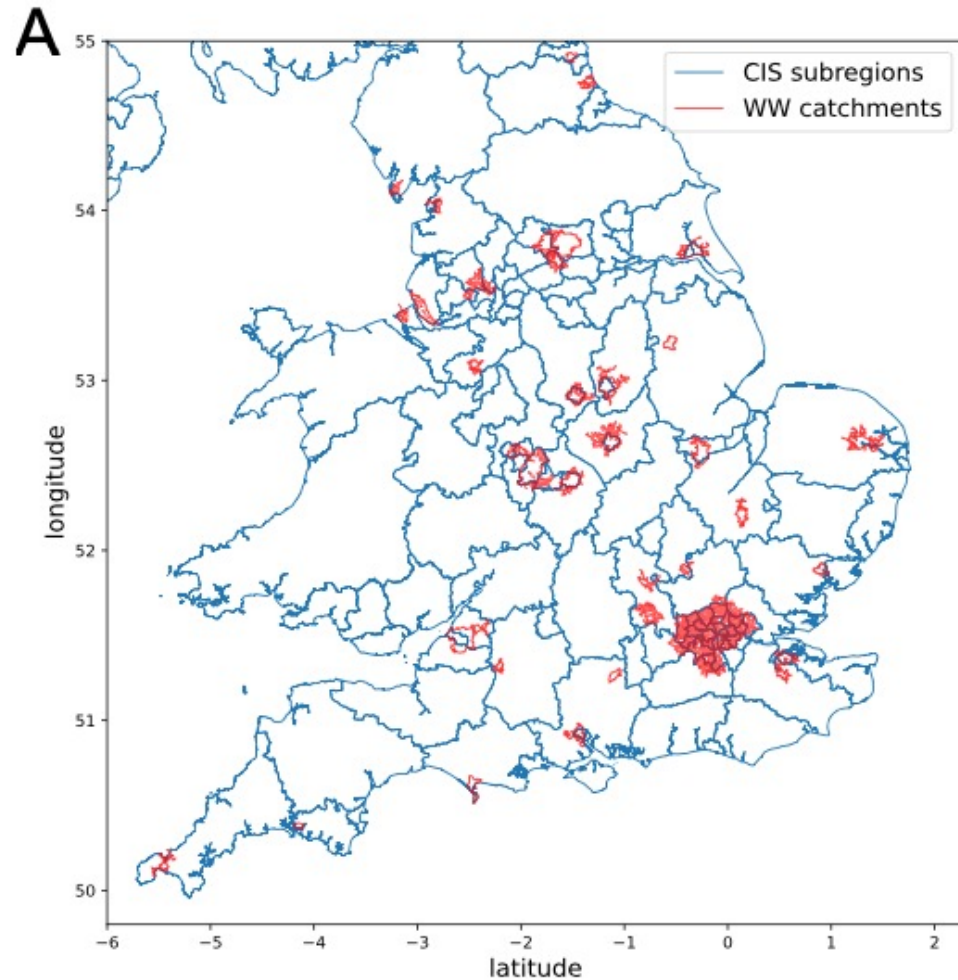
For accurate estimates of prevalence, need to know;

- Concentration in sample
- Wastewater flow (if it changes a lot)
- Shedding in stool (mean and var) \*
- Also be sure that virus shedding = being 'infected'

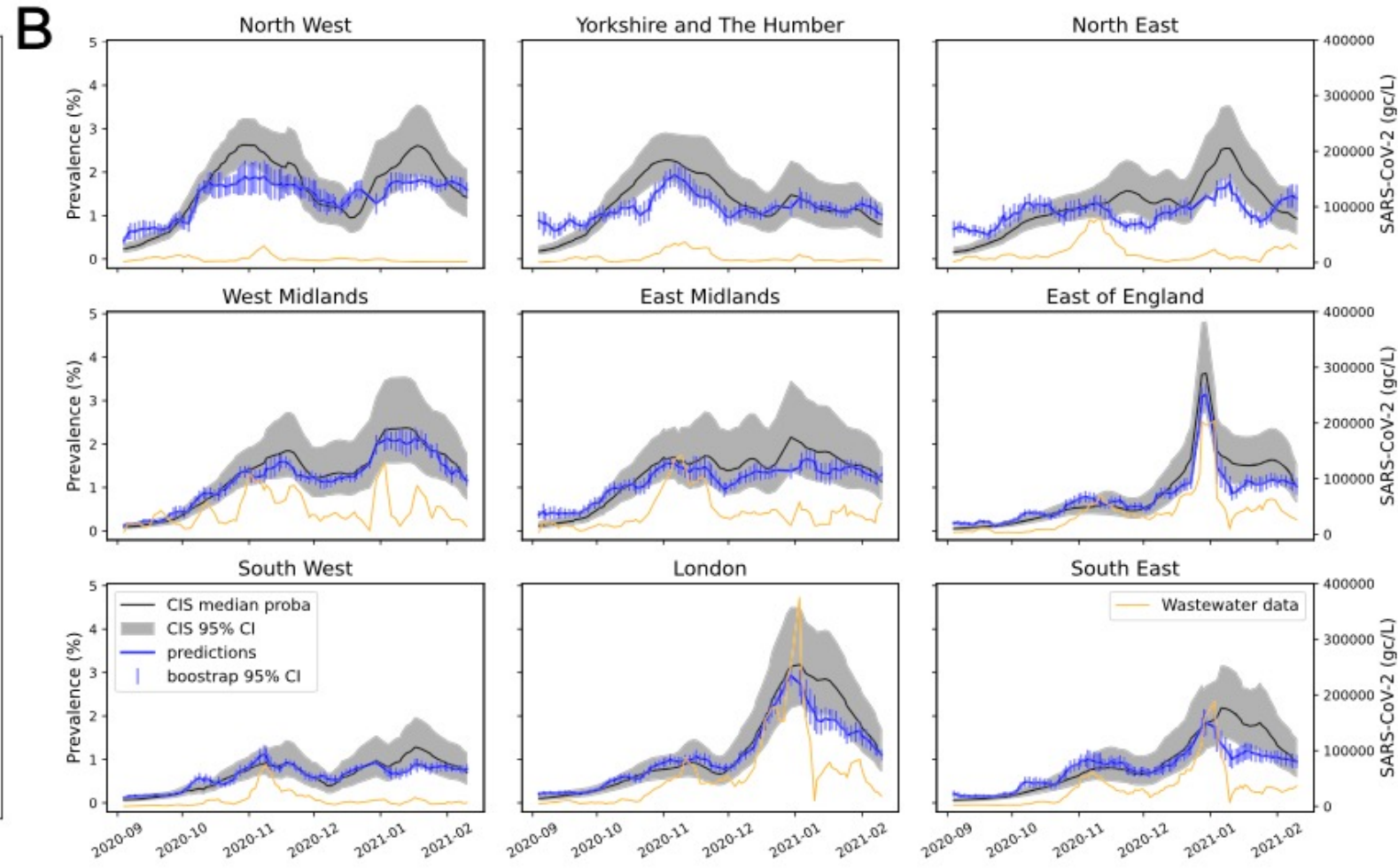




# Validation of Prevalence Estimates (England)



Locations of WW sites and comparisons to CIS regions



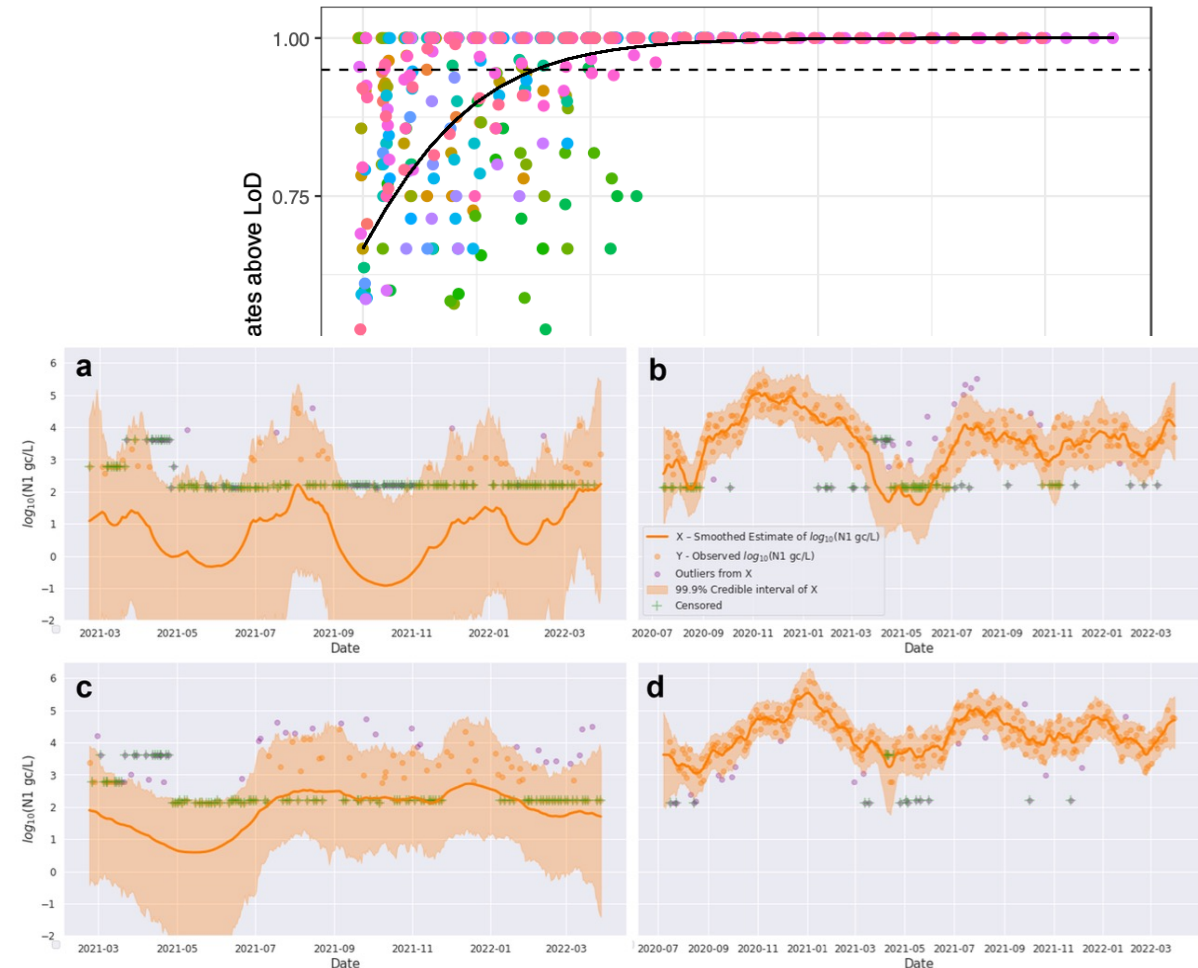
Blue: Estimates of prevalence from WW data & model  
Black & grey: CIS prevalence estimates  
Yellow: 'raw' estimate of WW data

**Morvan et al (2022)**

<https://doi.org/10.1038/s41467-022-31753-y>

# Messy Data...

- SARS-CoV-2 prevalence (via CIS) was often high
  - Proportion of ES samples >LOD also high
  - Some sites, and some periods had many below LOD
- Explored effect of 'left censoring'
  - **Lewis-Borrell et al (2023)**  
10.3934/math.2023859
  - Large effect in some sites
  - Data augmentation improved correspondence to CIS



Green: raw data (with LOD)  
Orange: Inferred gc/l measure



# Change in Pathogen Incidence

A measure of prevalence can be used to infer incidence,

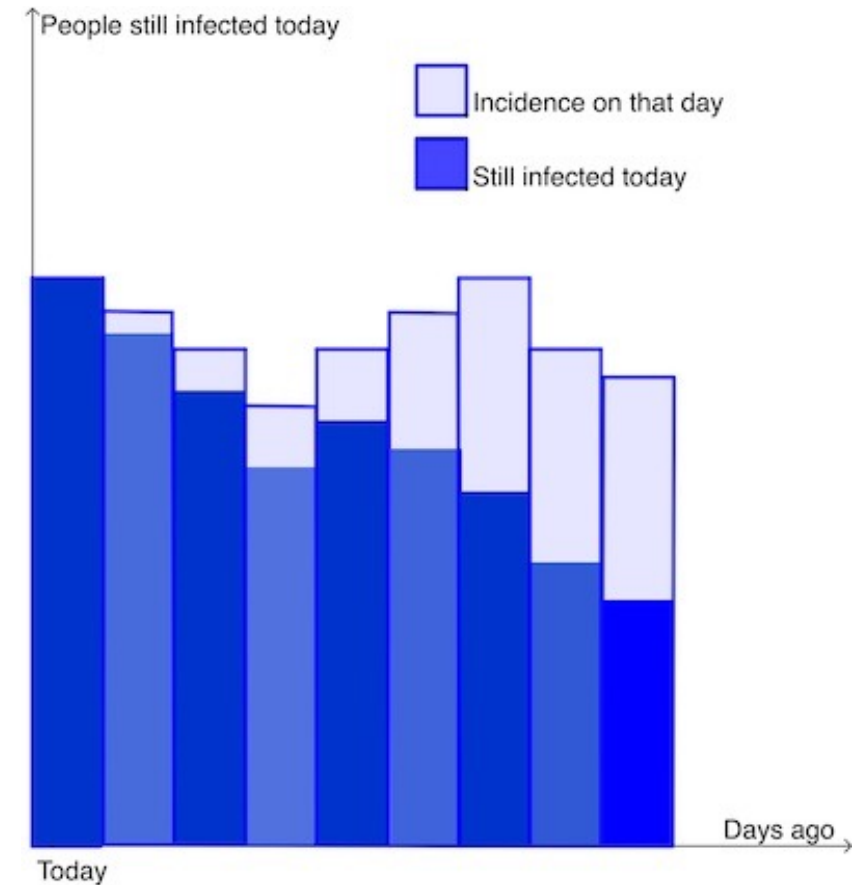
$$Prev(t) = \int_{T=0}^{\infty} Inc(t - T)Dur(T)dT$$

But we also want to account for,

- Changing shedding profile during course of infection

$$Shedding(t) = \int_{T=0}^{\infty} Inc(t - T)Shed(T)dT$$

- Deconvolution and then *usual methods to estimate Re*



<https://plus.maths.org/content/keeping-covid-19>  
(handy!)

# Validation of Incidence and Re Estimates (Switzerland)

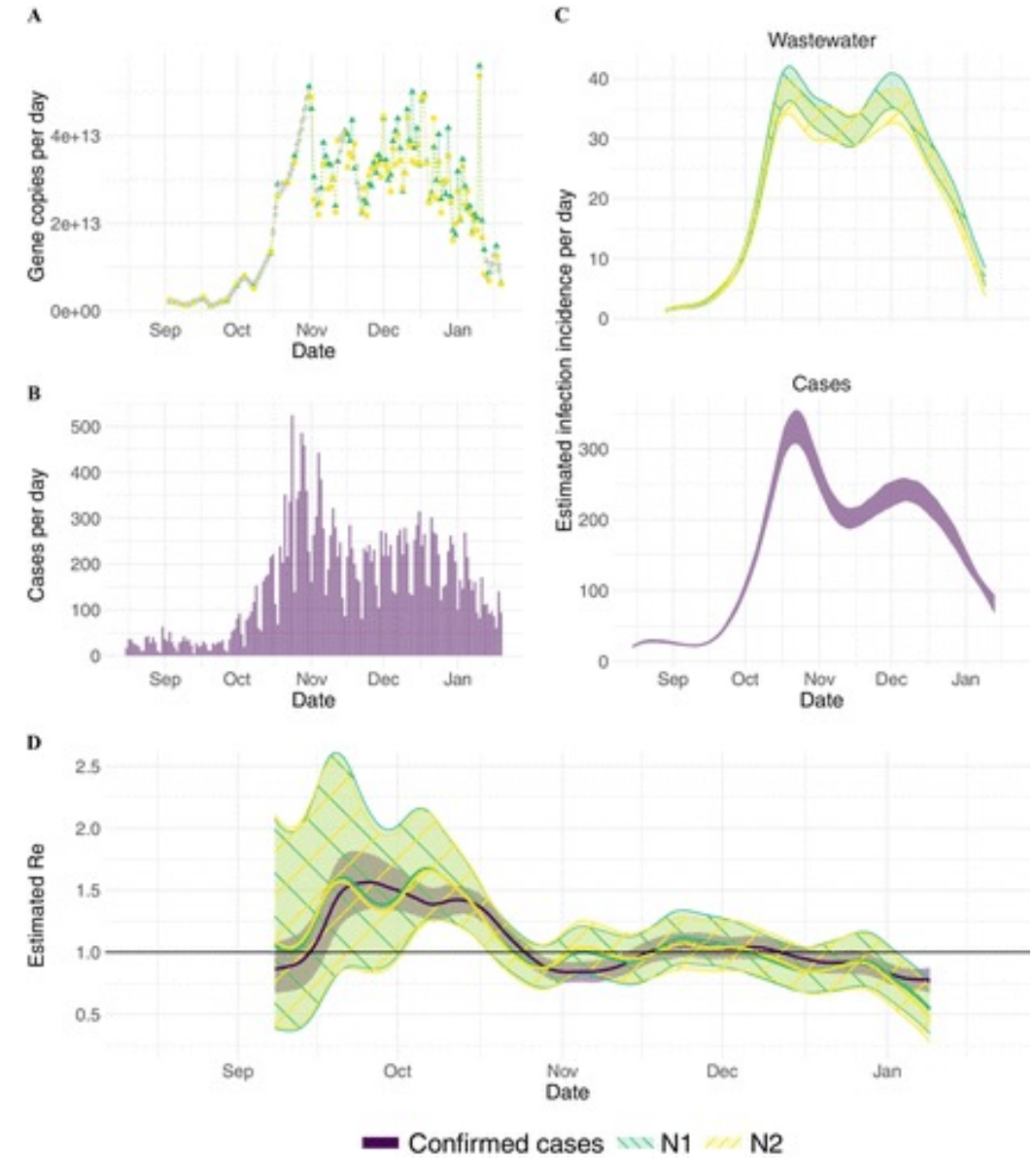
Described in full in **Huisman et al (2022)**

<https://doi.org/10.1289/EHP10050>

- Infection incidence inferred from WW and case data broadly agreed
- Ribbons of Re estimates had good overlap
  - More uncertainty from WW data

In addition to WW data (measure, flow),  
require shedding profile of the pathogen

- Some flexibility, depending on accuracy of output



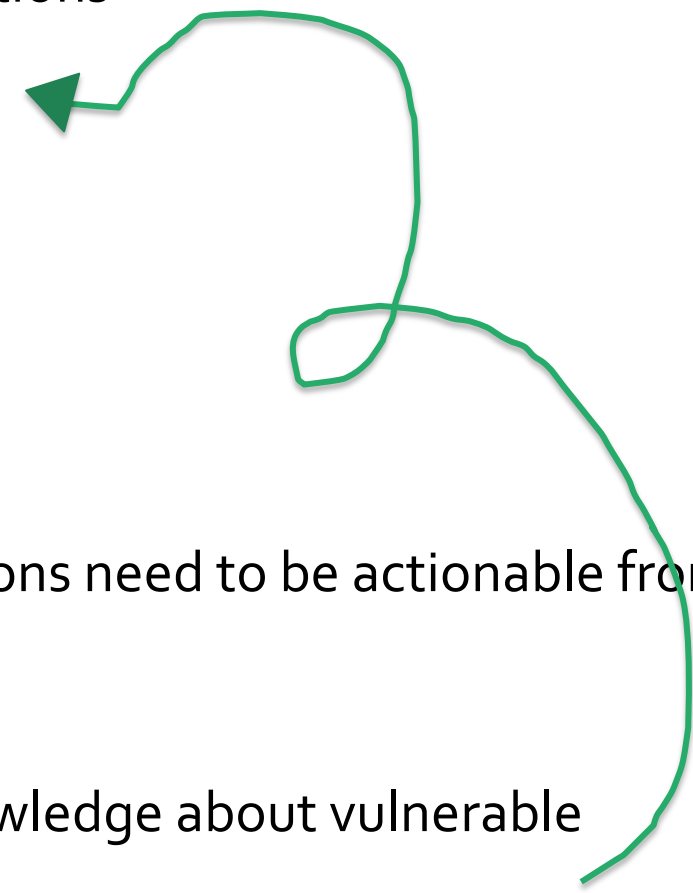
# Back to the Questions

	Pathogen measure	LOD effects	Shedding - average	Shedding - profile	Water flow
3. How has the prevalence of pathogen A changed in time?	X	X	X	-	X
4. How has the prevalence of pathogen A changed in time?	X	X	X	X	X
5. How does prevalence compare between locations?	X	X	X	X	XX

But also need to consider further questions:

- How representative is the ES sample of the population I'm interested in?
- What is the sensitivity / recovery of pathogen?
- Are inhibitors present in the sample?
- Has the sample been affected by degradation?

# So ES Data has Potential...

- Data collection that can be *designed* to answer specific questions
    - This hasn't happened much yet, but will happen soon
  - Sample collection is cheap!
    - ~£500/sample for lab costs
    - *Well, cheaper than CIS...*
  - Much ES data are publicly available (eg. dashboards)
    - Few issues of identifiability
  - For the data and analysis to be useful, clear uses / interventions need to be actionable from ES data
    - Are the data reliable enough?
    - Sampling the 'general population' means limited knowledge about vulnerable groups
    - What actions/interventions are appropriate?
- 

# From Research to Routine Use

- Increased interest in using ES since COVID-19
  - A very cross-disciplinary field
  - Evolving technologies
  - Evolving techniques (inc. analysis)
- Research needs for routine use **Shaw et al (2023)** doi/10.1038/s41591-023-02457-7:
  - Translation from HIC to LMICs – non-sewered settings
  - Designing a sampling scheme \*
  - Minimal criteria for reporting
  - Validation for new pathogens \*
  - Integration with clinical data \*
  - Compare costs of data collection to benefits of interventions \*
  - Best practice for communication

\* More  
modelling &  
analysis  
needed!



# Acknowledgements

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*And many more...*

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Slides will appear on  
[Github.com/kath-o-oreilly/presentations](https://github.com/kath-o-oreilly/presentations)