

Modelling hepatitis C among risk groups in Norway

- to monitor progress towards elimination

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Outline

1. Background: Hepatitis C in Norway, elimination goals
2. Model design, technical choices and challenges
3. Results and discussion

Background: Hepatitis C in Norway, elimination goals

WHO elimination goals for hepatitis C

- Chronic infection with hepatitis C virus (HCV) may cause progressive liver fibrosis
- A leading global cause of liver cirrhosis, cancer and death
- WHO estimates in 2019:
 - 58 million people lived with chronic hepatitis C
 - 1.5 million new infections/yr
 - 290 000 deaths/yr
- New oral direct-acting antivirals (DAA) are safe, effective and quick
 - Made free for everyone with HCV in Norway in February 2018
- Elimination goals by 2030:
 - 80 percent reduction in incidence from 2015
 - 5 new infections per 100 000 population, 2 per 100 000 for PWID
 - 2 deaths per 100 000
- NIPH has responsibility for monitoring progress towards elimination
 - Asked to make a model to estimate prevalence and incidence



Hepatitis C is concentrated in risk groups

- People who inject drugs (PWID) - active transmission
- Immigrants from high-prevalence countries - importers of chronic infection, little onward transmission (?)
- Other groups:
 - Persons infected through blood transfusion pre-1990
 - Men who have sex with men
 - Prevalence very low -> overshadowed by uncertainty in PWID + immigrants -> **not included**

Building on earlier work

Meijerink *et al.* *BMC Infectious Diseases* (2017) 17:541
DOI 10.1186/s12879-017-2631-2

BMC Infectious Diseases

RESEARCH ARTICLE

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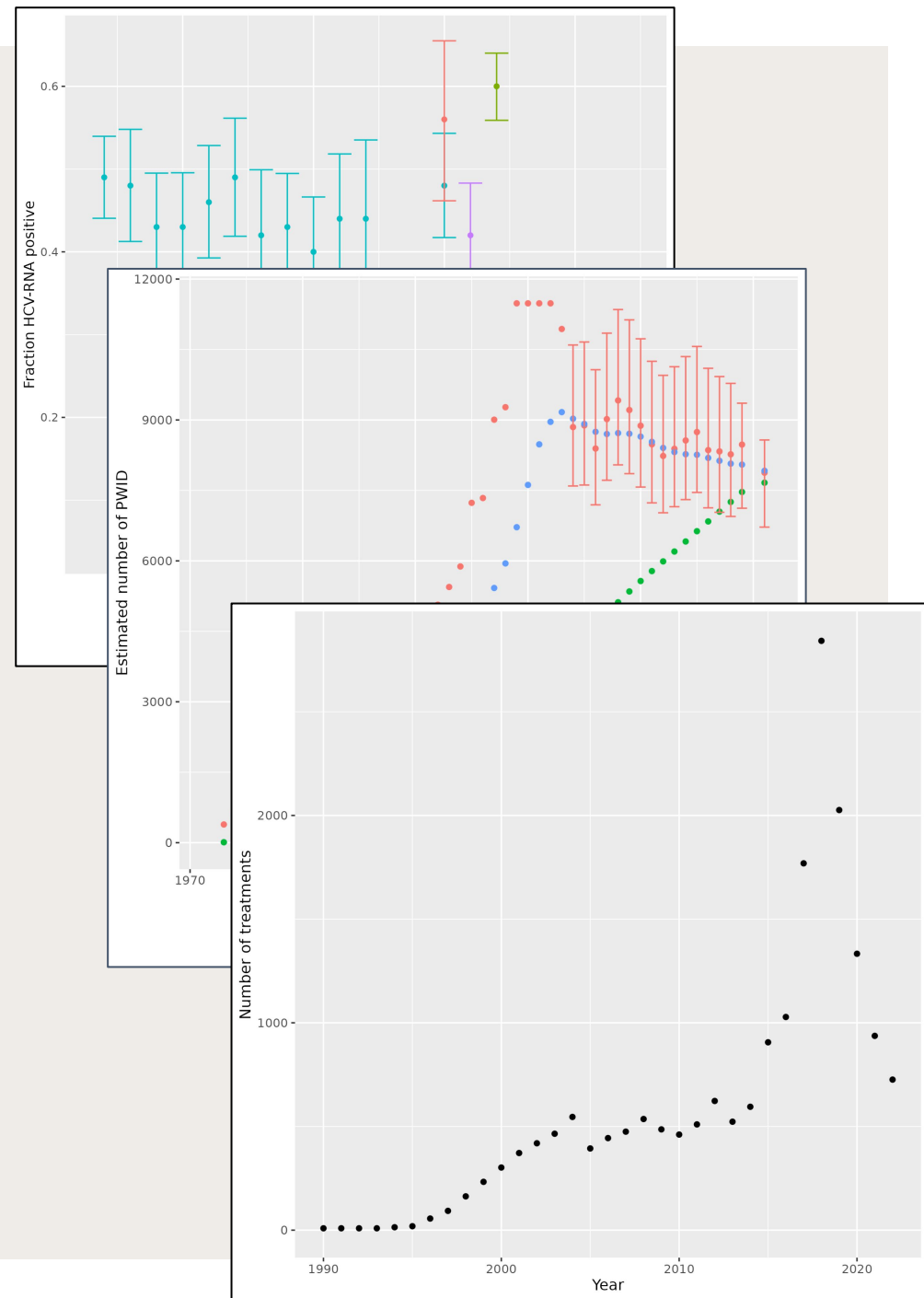
Modelling the burden of hepatitis C infection among people who inject drugs in Norway, 1973–2030

Hinta Meijerink^{1,2}, Richard A White¹, Astrid Løvlie¹, Birgitte Freiesleben de Blasio^{1,3}, Olav Dalgard^{4,5}, Ellen J. Amundsen¹, Espen Melum^{6,7,8} and Hilde Kløvstad^{1*}

- Modelling study on HCV among PWID in Norway
- 2017 - just before DAA treatments
- Main focus on severe outcomes
- Not straightforward to update and rerun their model, so we built a new one
 - However, many choices in our model inspired by this paper!

Patchy data sources

- RNA and antibody prevalence surveys - convenience sample among active PWID
- Total number of PWID - estimated from drug overdose deaths
- Number of treatments per year + estimated treatment success probability
- Immigrants: Lacking any kind of prevalence data for Norway



Model design

Data, technical choices & challenges

A stochastic compartmental model



- Inference is performed by a particle filtered Markov Chain Monte Carlo (pMCMC) engine implemented in the *mcstate (odin/dust)* framework within R
 - Parameters are a combination of random walk in time and time-constant, inferred jointly
- All data sources are evaluated together against all parameters in a total likelihood function

A stochastic

- Infection
- Carriers
- Frames
- All
- total

```
## Core equations for transitions between compartments:
update(S) <- S - n_SI
update(I) <- I + n_SI - n_IR
update(R) <- R + n_IR

## Individual probabilities of transition:
p_SI <- 1 - exp(-beta * I / N) # S to I
p_IR <- 1 - exp(-gamma) # I to R

## Draws from binomial distributions for numbers changing between
## compartments:
n_SI <- rbinom(S, p_SI)
n_IR <- rbinom(I, p_IR)

## Total population size
N <- S + I + R

## Initial states:
initial(S) <- S_ini
initial(I) <- I_ini
initial(R) <- 0

## User defined parameters - default in parentheses:
S_ini <- user(1000)
I_ini <- user(1)
beta <- user(0.2)
gamma <- user(0.1)
```



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README.md

<https://github.com/Gulfa/metapop>

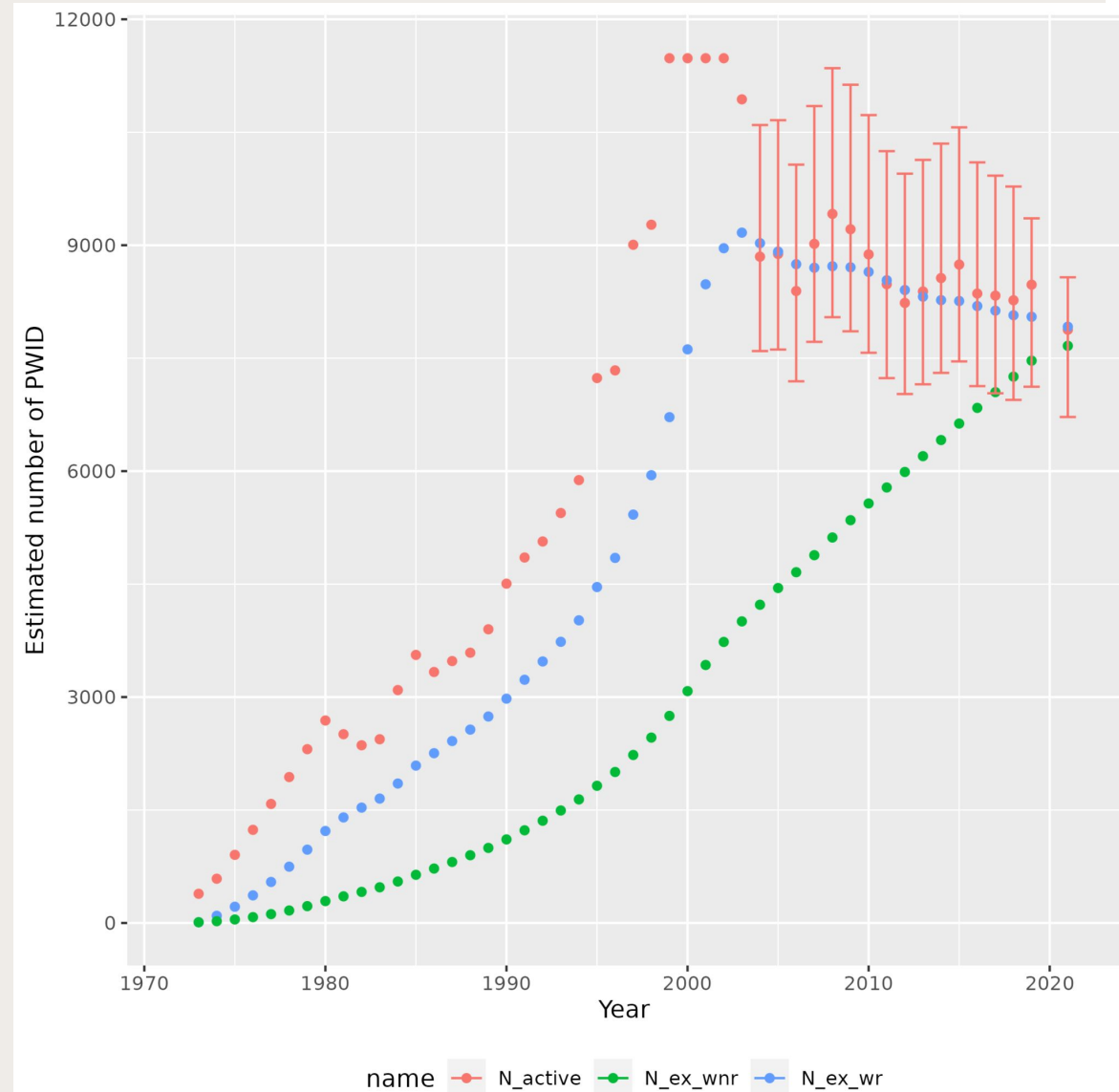
metapop

A stochastic metaopulation model with vaccination and multiple strains. Initially developed for modelling COVID-19 in Norway



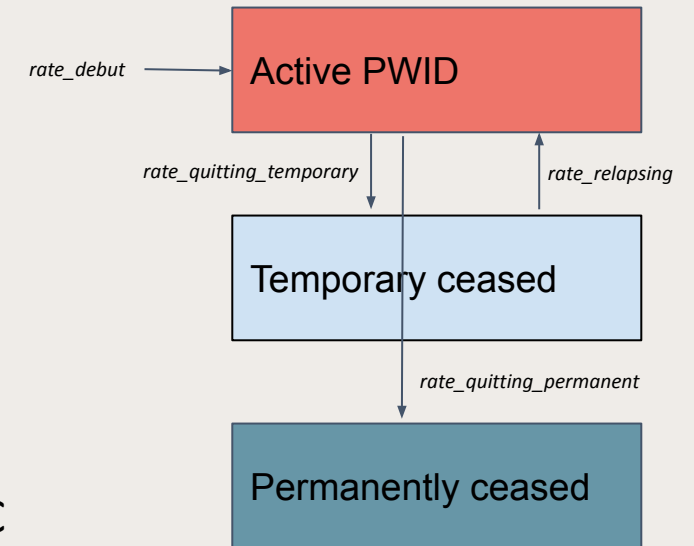
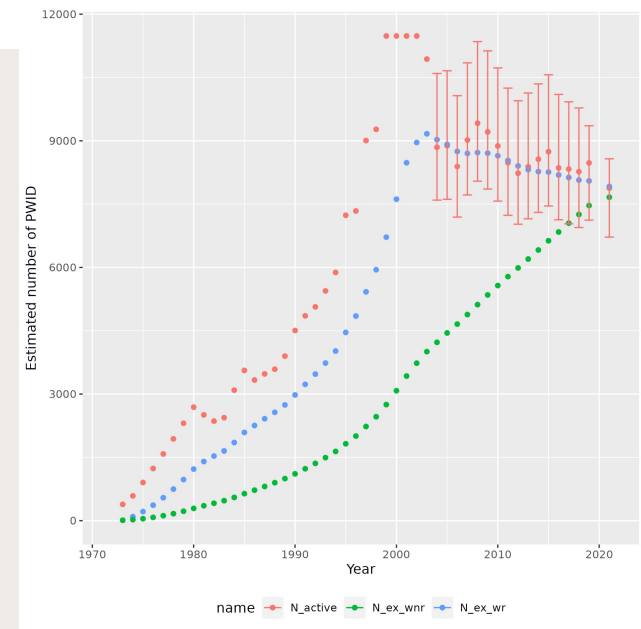
The PWID population

Estimated by mortality
multiplier method courtesy
of NIPH, Dept. of Alcohol,
Tobacco and Drugs



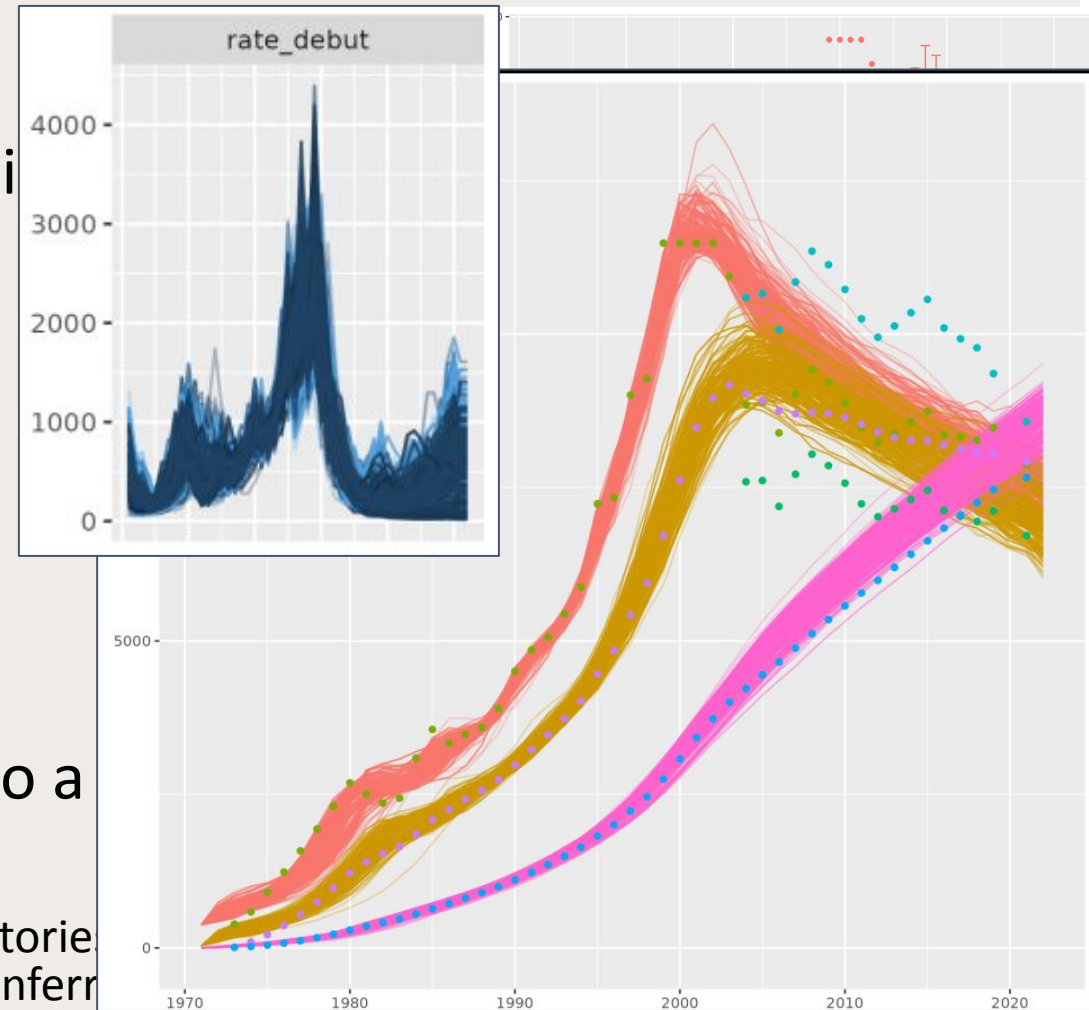
The PWID population

- We make a model for the PWID population consisting of three groups
 - Active PWID - currently injecting
 - PWIDs temporarily ceased injecting, will relapse
 - PWIDs permanently ceased injecting
- Parametrise this by four rates
 - *rate_debut*
 - *rate_quitting_temporary*
 - *rate_relapsing*
 - *rate_quitting_permanent*
- To give model necessary flexibility, we do a random walk on *rate_debut* over time
 - The particle filter is used to filter the desired trajectories
 - The 3 other parameters are kept constant in time, inferred by MCMC



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Sequential Monte Carlo (particle filter)

At each time step, the model draws a random perturbation on *rate_debut* (and on *rate_treatment*)

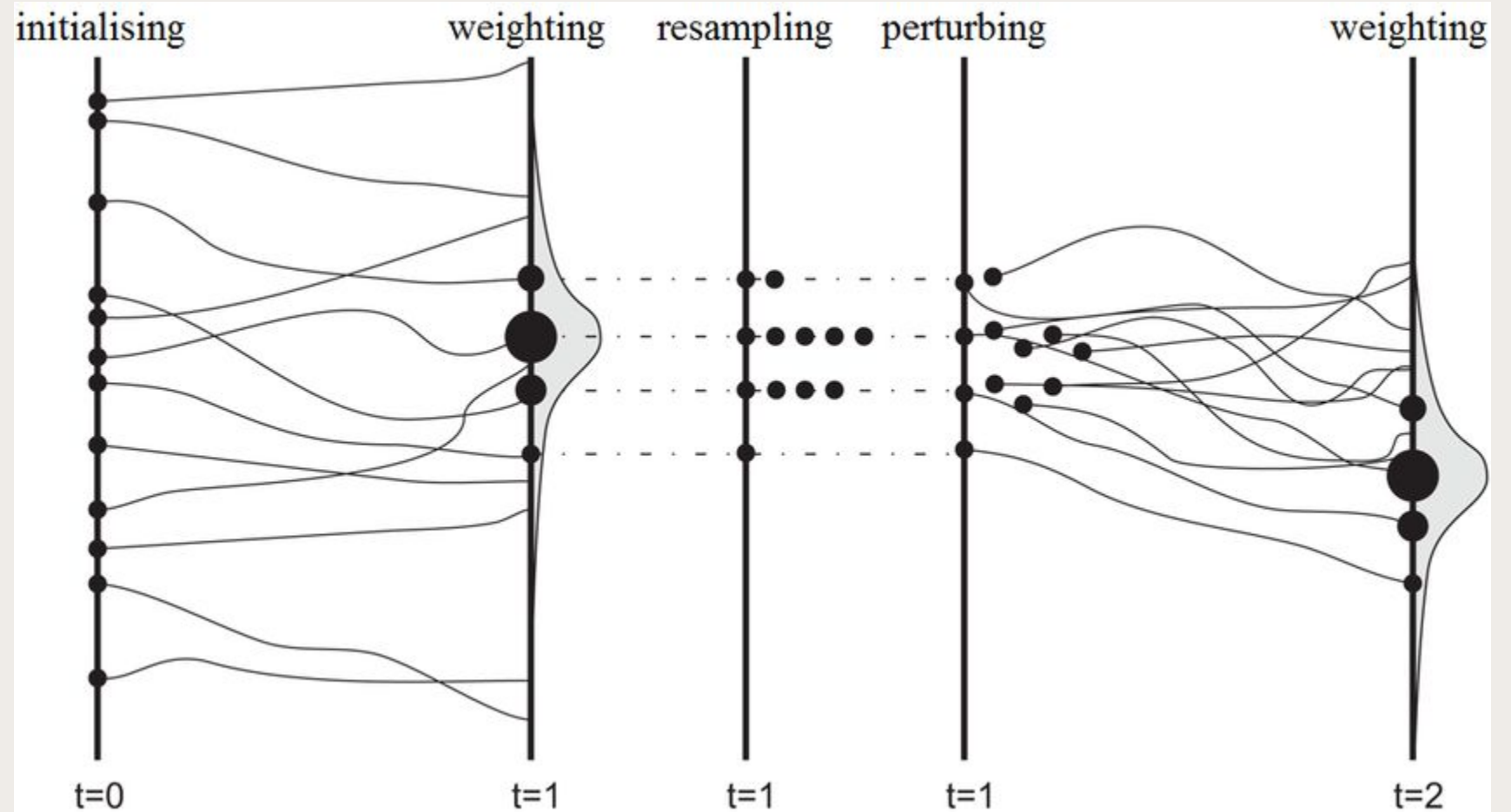
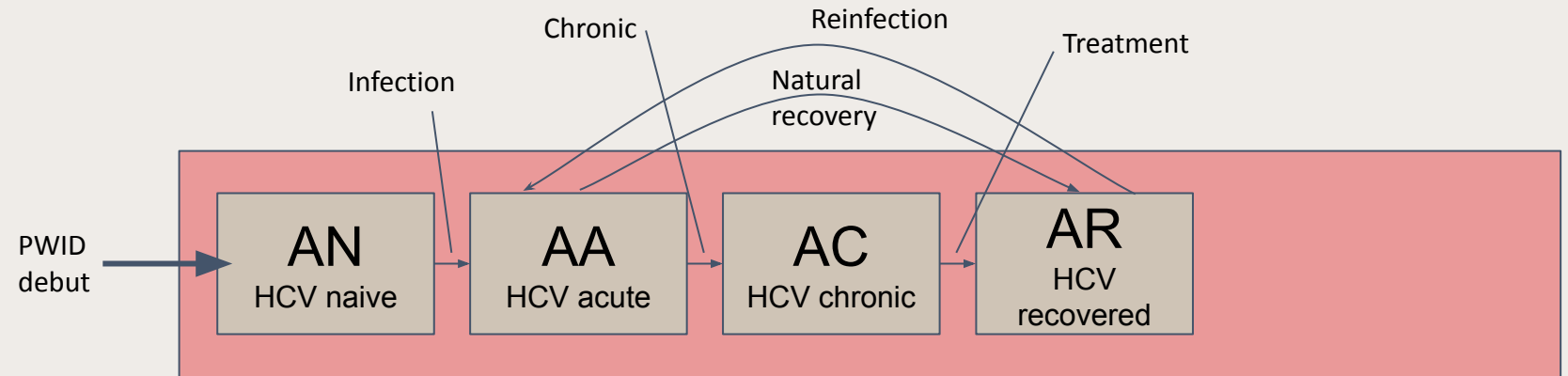
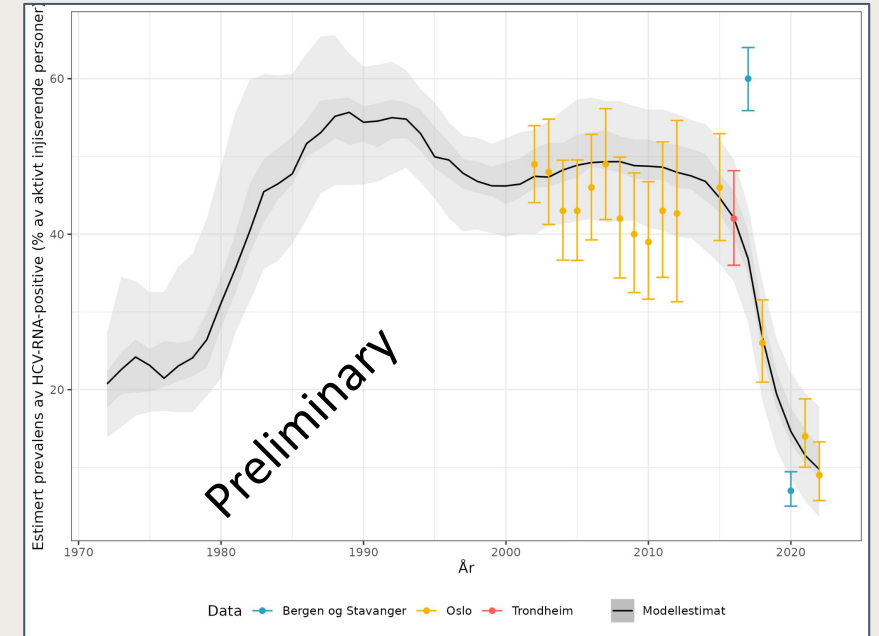


Illustration by Danilo Alvares

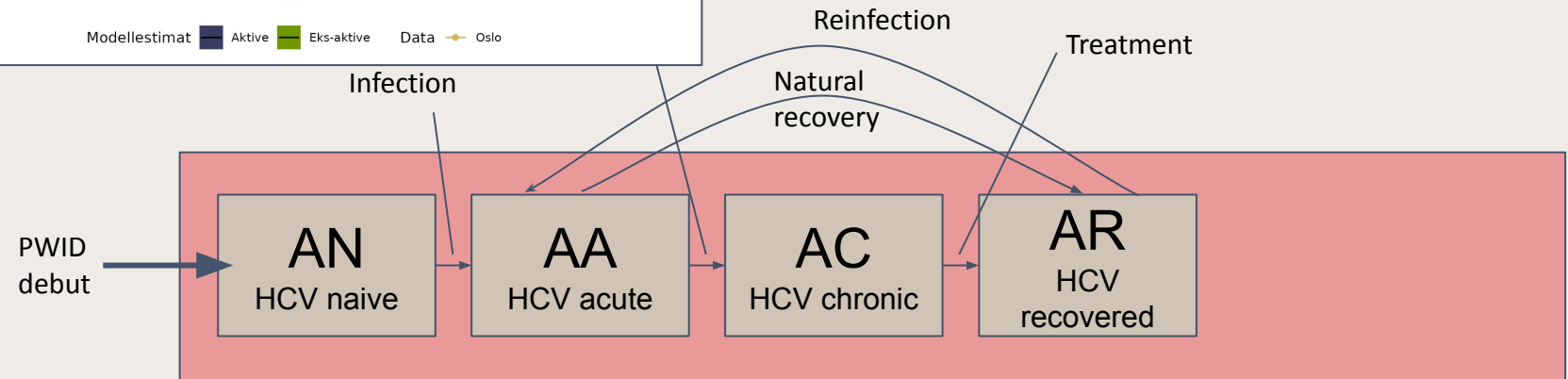
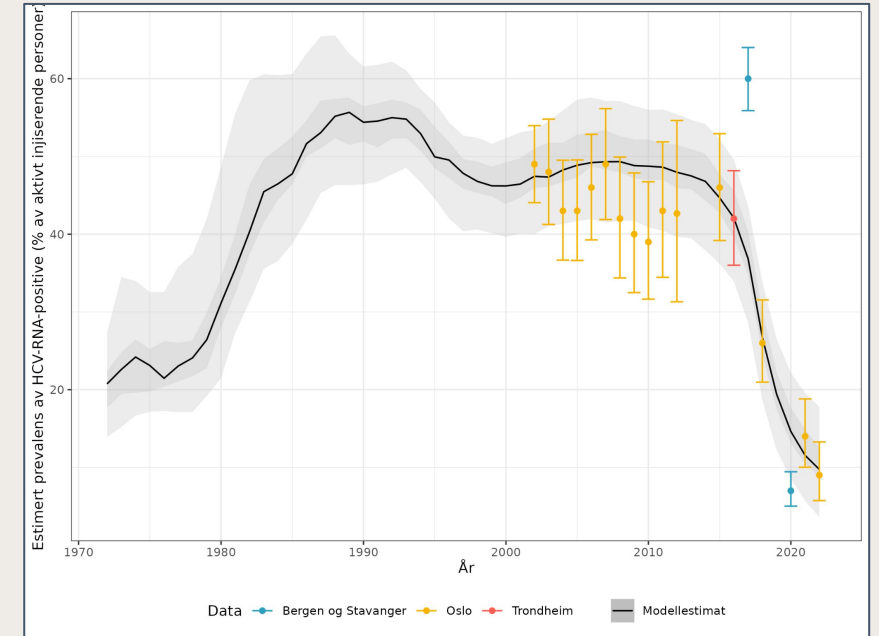
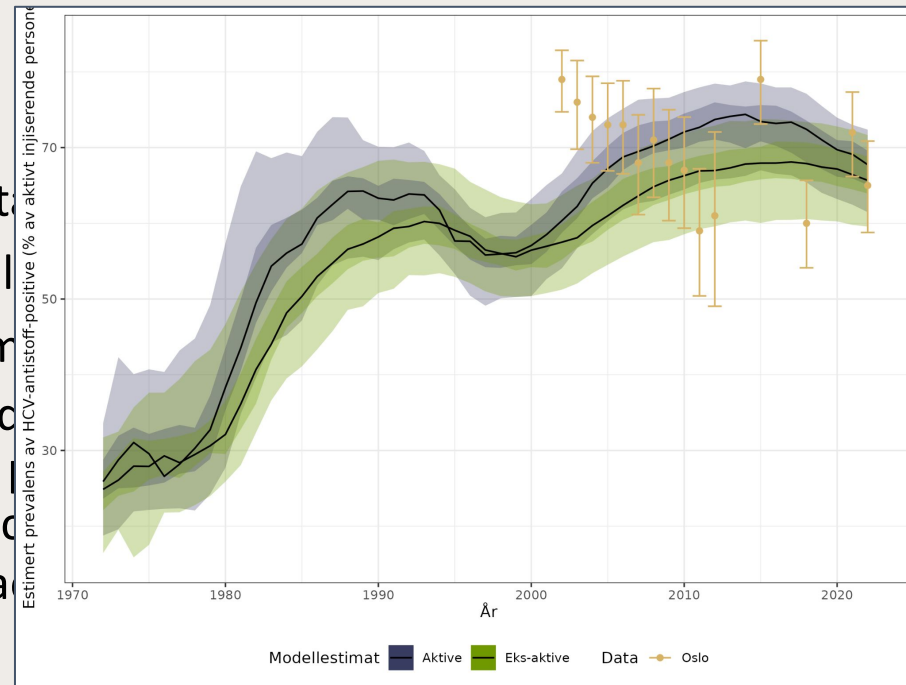
Spreading of HCV among active PWID

- Assume a constant underlying transmission rate β through simulation, inferred by MCMC
- Effective transmission rate lowered gradually by
 - needle and syringe and opioid substitution treatment programmes
 - geographical dispersion - GINI coefficient
- Informed by fraction active PWID who are RNA positive



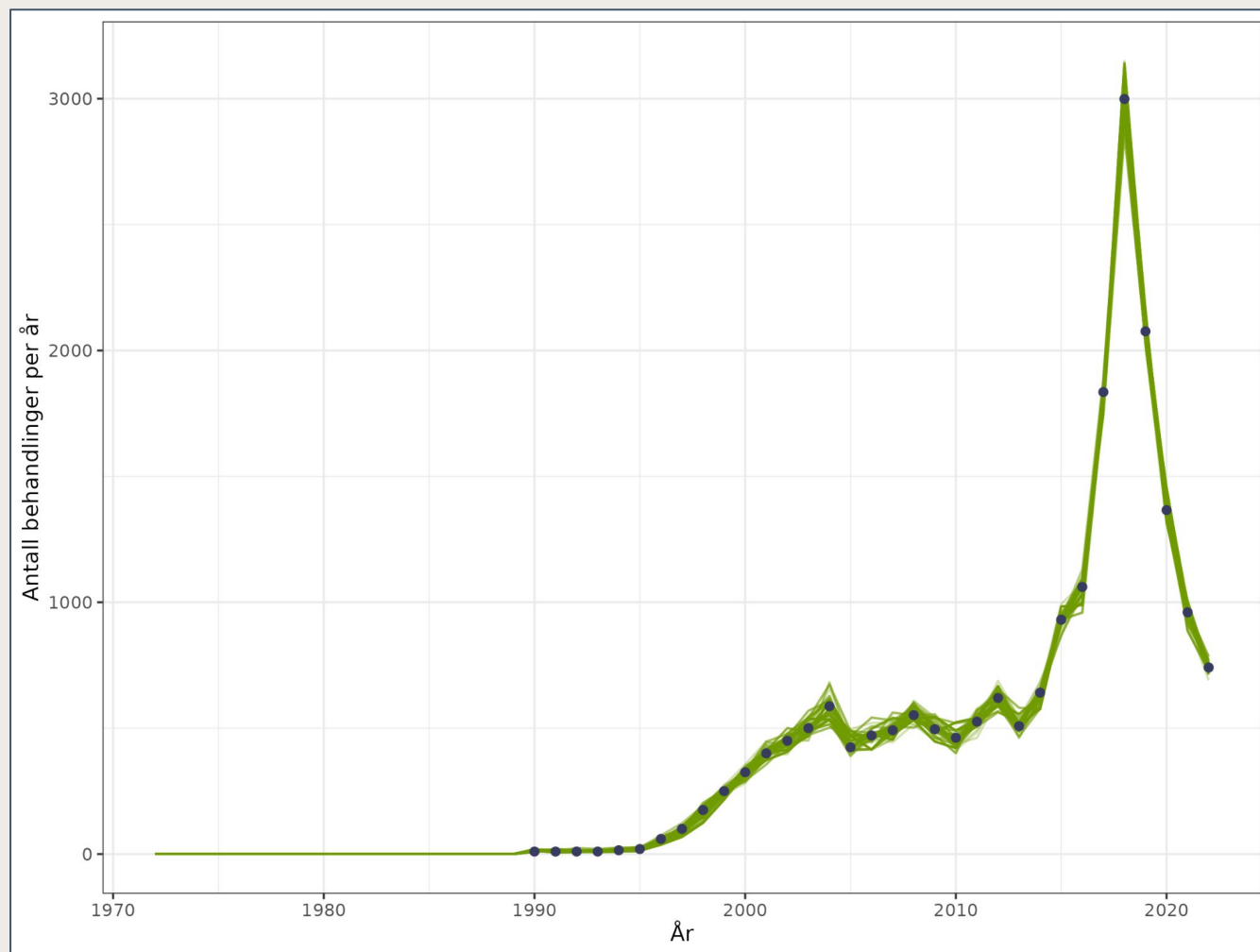
Spreading of HCV among active PWID

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- Effective transmission
 - Needle and syringe sharing
 - Treatment
 - Geographic mobility
- Informed by fraction of PWID who are HCV positive



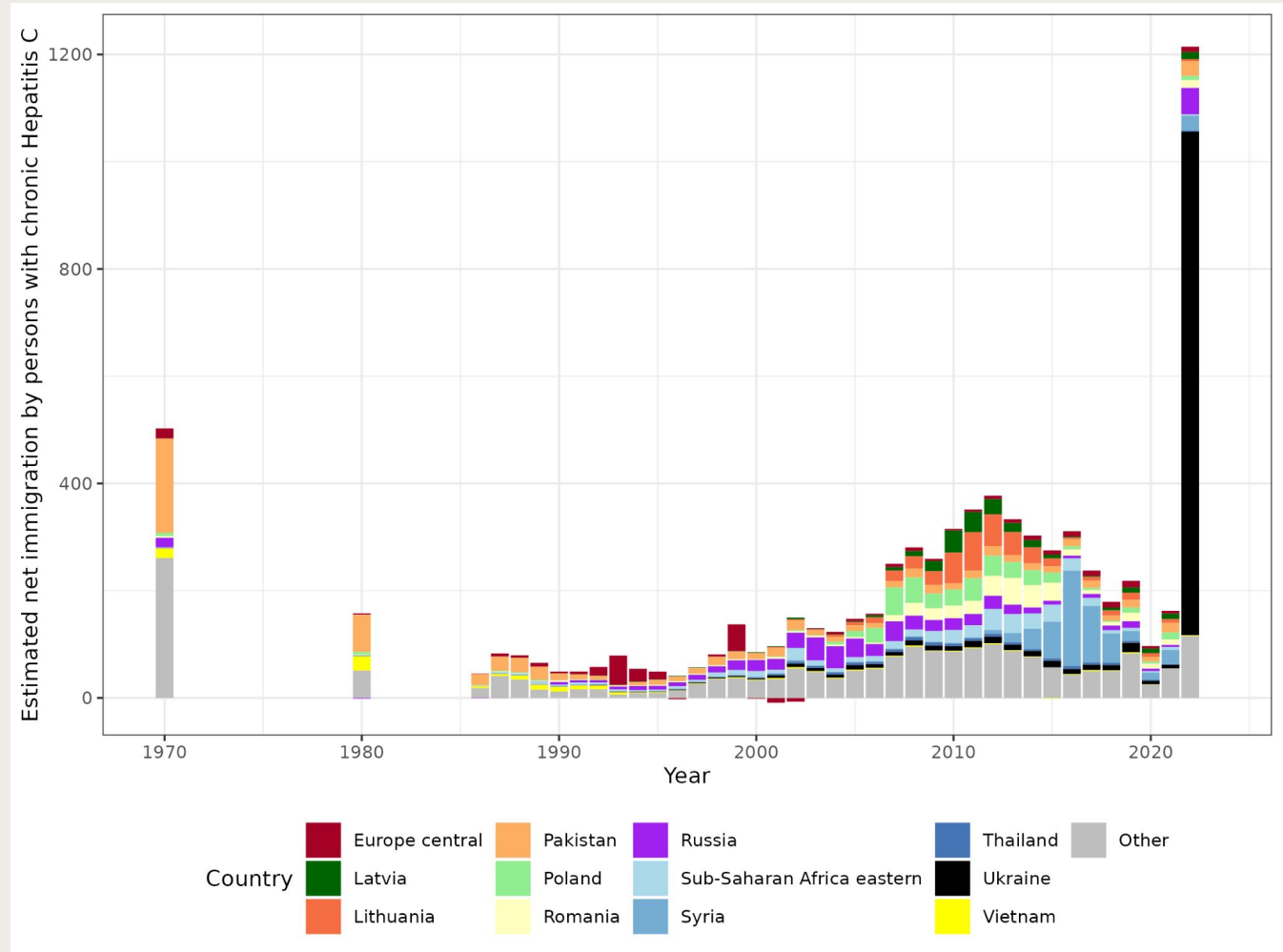
Treatments

- Treatment rate varies significantly from year to year, and thus is modelled as another random walk parameter against the particle filter.
- We assume that all groups (active and ex PWID, immigrants) have the same per-person probability of seeking treatment



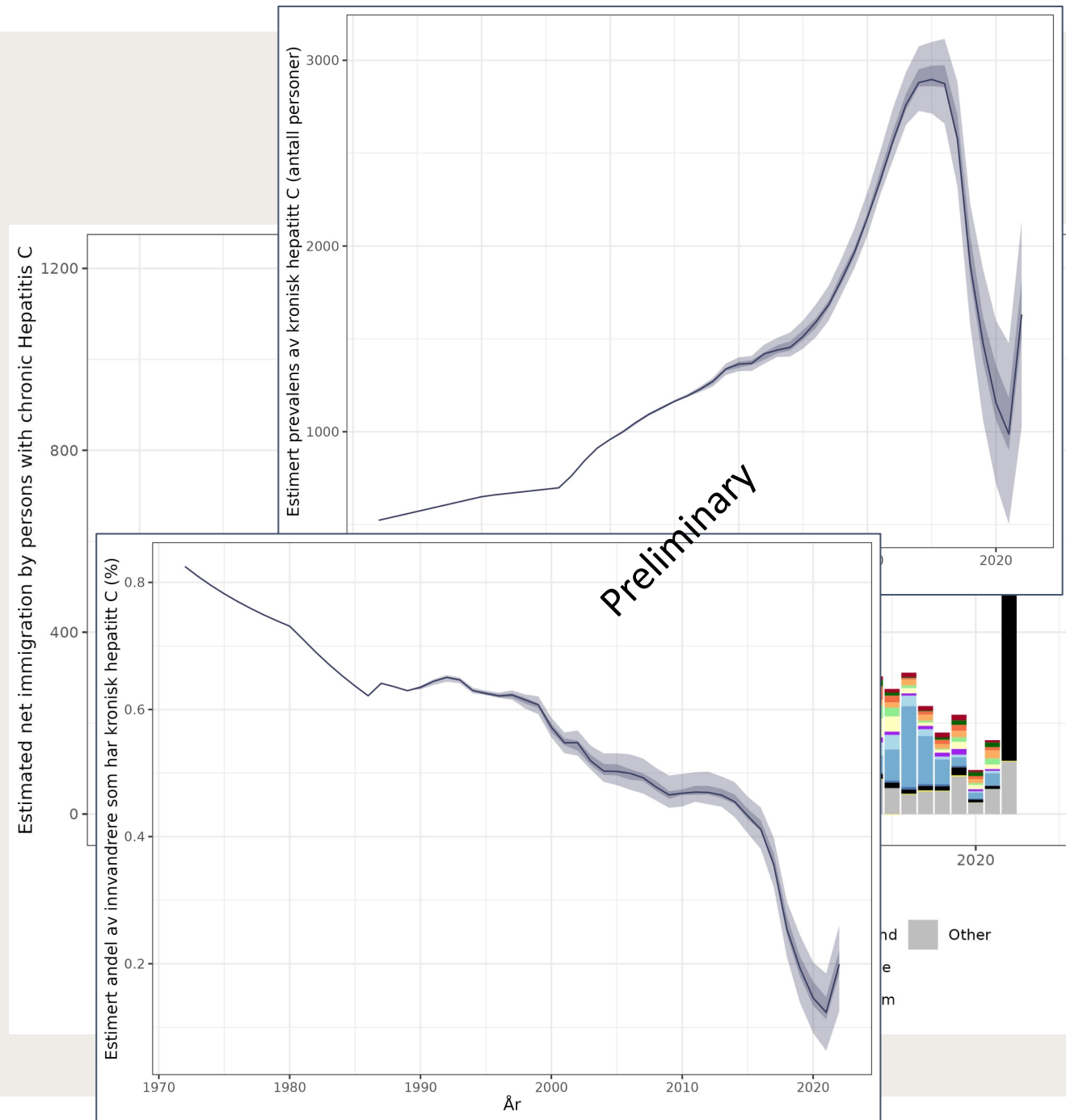
Immigrants

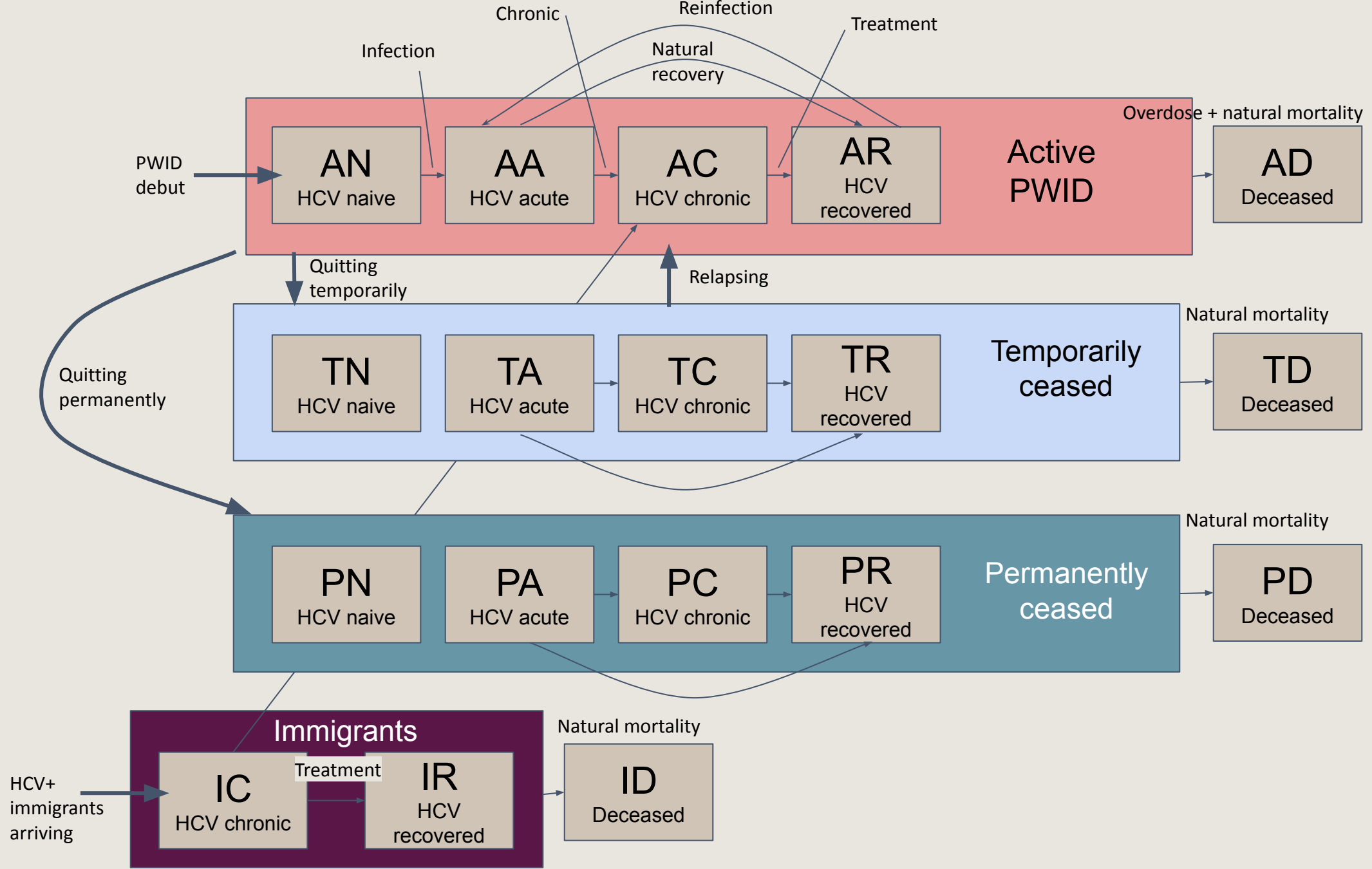
- We have no data on Norwegian immigrant HCV prevalence
- Immigrants are however needed in the model because they couple to the PWID through treatment data
- We make a simplistic assumption:
 - yearly immigration data +
 - per-country HCV prevalence estimates



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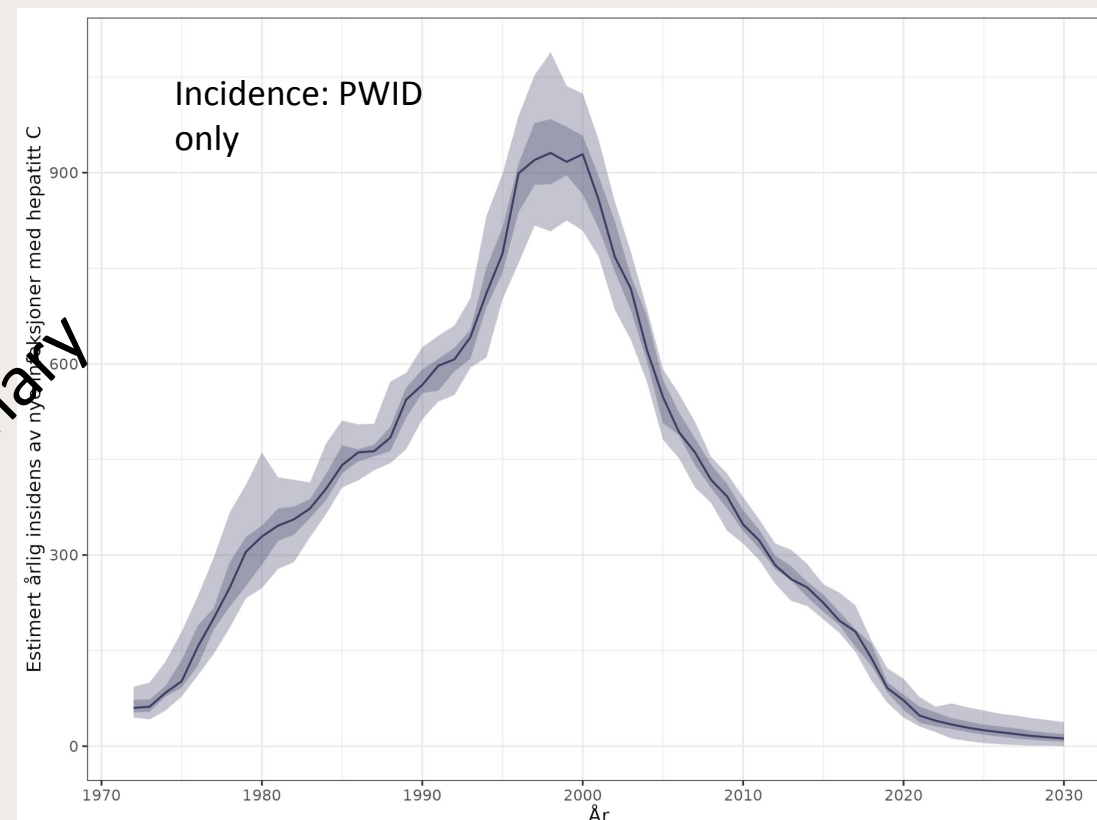
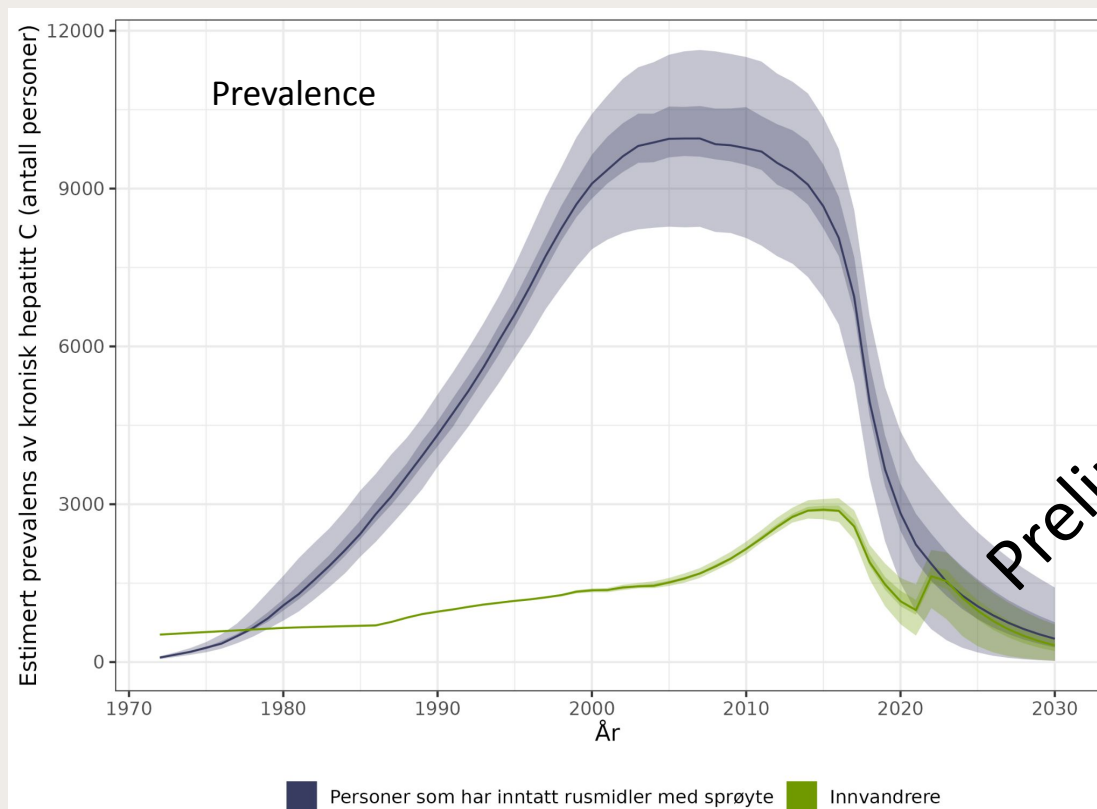




Results and conclusions

The model suggests Norway is on track to reach elimination goals

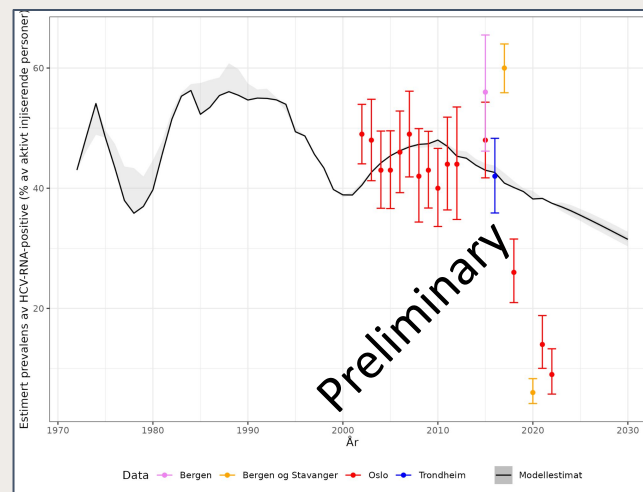
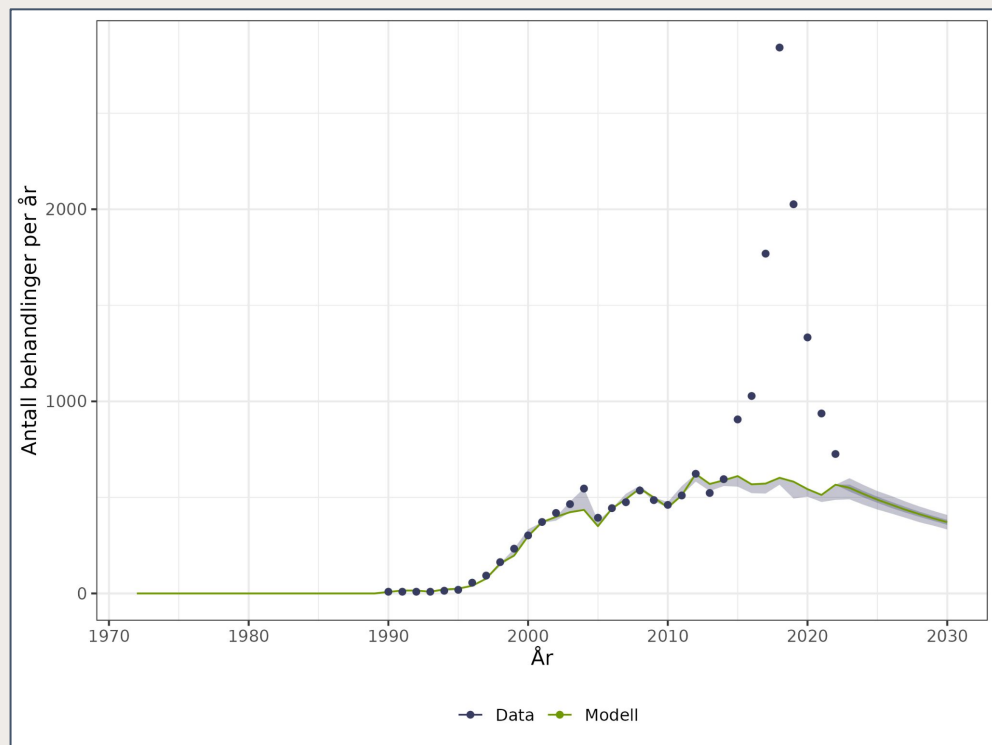
The model predicts forward in time by keeping random-walk parameters constant at last value



The large error bands make the case for closer monitoring of risk groups

Counterfactual: The impact of DAA treatments

Keep the transmission rate fixed at with-treatment model estimate,
keep treatments at 2013 level



Conclusions and outlook

- The hepatitis C prevalence & incidence is estimated to be rapidly declining, and this is largely attributable to treatment
- Norway is estimated to be **on track** to reach the elimination goals, and the hepatitis C epidemic among PWID is predicted to die out
- More sensitivity analyses needed, e.g.
 - Vary assumptions on immigrant prevalence
 - Test assumption of equal treatment uptake
- Question: How to choose the “correct” number of particles in the particle filter to represent uncertainty and not overfit? (Presently 10^3 - 10^4)

This work is done together with **Robert Neil Whittaker**, with valuable input from Hilde Kløvstad, Sasi Kandula, Gunnar Rø, Birgitte de Blasio and others.