

Community viral load

We explore how sexual behaviour and STI co-infection (HSV2, HCV) dynamics jointly influence the evolution of HIV viral load, and subsequently the rate of new HIV infections.

1. Implement HIV-HSV2 (resp. HCV) co-infection in SIMPACT
2. Simulate
 - ~ 10,000 scenarios in a homosexual population of 10,000 men:
 - introducing ART in 1994 (simulate as an intervention event, adapt every 1-2 year, vary `monitoring.cd4.threshold`);
 - and varying the following parameters:
 - 1) parameters related to sexual behavior (agegap formation hazard, MSM version):
 - $\alpha_{numrel,sum}$: penalty per ongoing relationship
 - $\alpha_{numrel,diff}$: penalty for differing degrees (very negative value results in more assortative mixing w.r.t. degree)
 - $\alpha_{eagerness,diff}$: penalty for differing intrinsic sexual drive (very negative value results in more assortative mixing w.r.t. intrinsic sexual drive)
 - scale parameters of the distribution from which individuals draw intrinsic sexual drive value (E_{man1} and E_{man2}). This scale parameters influences the amount of heterogeneity in intrinsic sexual drive levels in the population. Simulate different types of distributions.
 - 2) parameters related to intra-couple HIV transmission
 - rho: correlation between SPVLs of infector and newly infected individual (to study SPVL heritability)
 - d: penalty on transmission hazard per ongoing relationship
 - mean and stdev of the distribution from which SPVL values are drawn for individuals who acquire infection in the course of a simulation run
 - mean and stdev from which SPVL values are drawn (investigate role of founder effects).

We can sample using Latin Hypercube Sampling. Software: R – lhs package, function `randomLHS`.

Key assumptions:

- heritability of set point viral load
 - agegap formation hazard
3. For each of the 10,000 scenarios, we calculate the following summary statistics:
 - 1) To characterize sexual network connectivity
 - point prevalence of concurrency at t15 (and other time points)
 - summary statistics of network connectivity:
 - degree distribution:

For a sexual network, vertices represent persons and edges sexual

relationships between them. The degree d of a vertex represents the number of sexual partners of a person, and the degree distribution is the distribution of d .

- conditional distribution: relative frequencies of the form $f_{d'|d}$ with which edges from a vertex of degree d are connected to another vertex of degree d' . Means of these conditional distributions as a function of d summarize associations between persons with a high number of sexual partners and persons with a low number of partners.

- degree likelihood: measure proportional to the sum, over all edges, of the products $d(v_1)d(v_2)$ of the degrees of the vertices v_1, v_2 defining the edges $e = (v_1, v_2)$

- closeness centrality: high closeness centrality vertices correspond to highly influential individuals. People with high closeness centrality in a sexual network are highly responsible for the spread of HIV.

- betweenness centrality: indicates the extent to which individuals transmit HIV to others within the sexual network.

- average distances between distinct vertices: can be used to describe geographical distances between persons.

- summary statistics of network assortativity:
 - assortativity coefficient: minimum (between -1 and 1) = perfectly dissortative, every edge connects vertices (individuals) of two different categories; 1 = perfectly assortative, edges connect only vertices (individuals) from the same category; 0 = random.
 - for continuous characteristics (e.g. age): Pearson correlation coefficient.
- summary statistics of network heterogeneity:
 - degree distribution

2) To characterize time-trend in viral load distribution

- mean and median \log_{10} SPVL of seed HIV infections
- difference $t_0 - t_{15}$ in mean, median, stdev and IQR \log_{10} viral load
- difference $t_0 - t_{15}$ in mean, median, stdev and IQR \log_{10} SPVL
- also other time points than t_{15}

3) To characterize the course of the HIV epidemic

- HIV prevalence at t_{15}

Commented [DH1]: Is dus iets anders dan dat we dachten tijdens onze meeting op 9 nov. Is dit wel zinvol om dit te berekenen?

Commented [DH2]: Heeft volgens mij niet echt veel te betekenen in termen van sexual networks. Weglaten?

- Cumulative HIV incidence
 - Average age among HIV infected people at t15
 - Also other time points than t15
- 4) To characterize transmission dynamics
- Average number of new infections over the complete infectious period
 - Average number of new infections over the complete infectious period, per log10 SPVL bin
- 5) To characterize HIV-HSV2 (resp. HIV-HCV) co-infection
- HSV2 (resp. HCV) prevalence at t 15
 - HIV-HSV2 (resp. HIV-HCV) prevalence at t15
 - Cumulative HSV2 (resp. HCV) incidence
 - Cumulative HIV-HSV2 (resp. HIV-HCV) incidence
 - Average age among HSV2 (resp. HCV) infected people
 - Average age among HIV-HSV2 (resp. HIV-HCV) infected people
 - Also other time points than t15
4. Symbolic regression models (surrogate models) will be fitted to evaluate links between the summary statistics above. Software: Java, UA (code developed by Lander Willem & Sean Stijven); R: rgp package + eventually multivator package (R).
5. Sensitivity analysis:
- examine the role of founder effects (extremely high / low VL in seeding HIV infections) in driving the time-trend in community VL distribution;
 - study the effect of relaxing the assumption that intra-couple transmission rate is independent of the number of ongoing relationships individuals have.
- Therefore, consider the following two models
- 1) intra-couple transmission is a function of the viral load

$$h = \exp(a + b \cdot V^{(-c)})$$
 where V is the viral load.
 - 2) Sexual activity within relationships may decrease with growing numbers of concurrent relationships that man 1 (P_i) and man 2 (P_j) have.

$$h = \exp(a + b \cdot V^{(-c)} + d \cdot (P_i + P_j))$$