

Modeling Drug Use in Communities

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Introduction

For years, policymakers have been asking why and how do drugs travel through the community and what can they do to work minimise the harm they potentially cause. In this paper, we aim to understand how drug use and addiction develops in a community and what factors can be tweaked to alter the short term and long term distributions of drug use and addiction in the community.

// Add in past research

Method

Initial Model: SIR

It's common to liken drug use to disease, treating drug addiction as an infection that can be recovered from. Using this analogy, we began looking into how diseases are modeled and how we can adapt this model to fit drug usage. One of the simplest drug models is the SIR model, a compartment model based on the concept of mass action that sees infections as caused by interactions between infected people and susceptible people. There are a few reasons that this model is good fit to begin modeling drug use: Firstly, this model is very simple and easily adaptable to extra compartments or conditions on transfer. Secondly, since we are looking at drug use in a population, relying on mass action rather than discrete modeling allows us to simplify the model. Thirdly, the idea that addiction is caused by individuals interacting seems to be a good assumption for drug use as you would expect people to become addicted by being around other people with the drug.

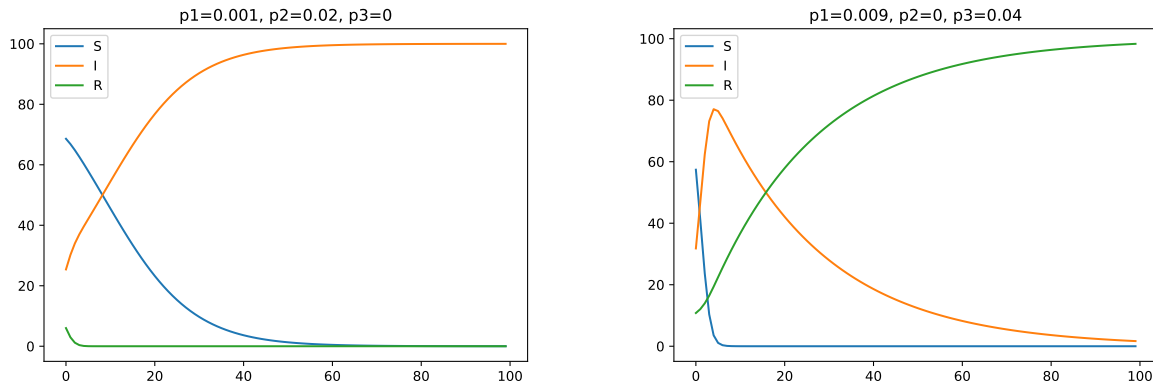
One divergence we made from the most basic SIR model is allowing for relapse from the recovered group back into the infected group. This represents how recovering addicts will become addicted again at a different rate to people who have never used the drug before. However, we never move people back into the susceptible group as addiction is said to be 'lifelong'.

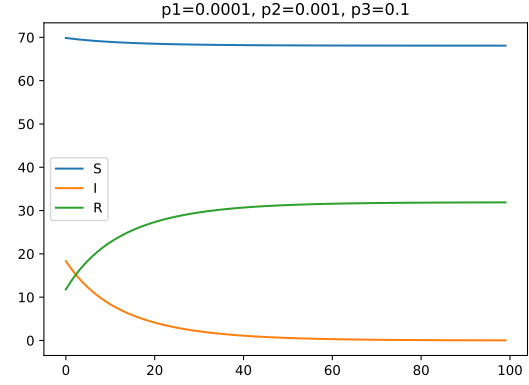
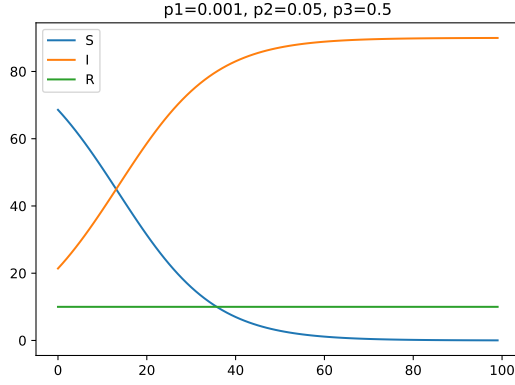
The SIR model can be defined with the following equations:

$$\begin{aligned}\frac{dS}{dt} &= -\pi_1 * S * I \\ \frac{dI}{dt} &= \pi_1 * S * I + \pi_2 * R * I - \pi_3 * I \\ \frac{dR}{dt} &= -\pi_2 * R * I + \pi_3 * I\end{aligned}$$

With the parameters π_1, π_2 and π_3 representing the 'infectiousness' of the drug for susceptible people, the infectiousness for recovering people and the rate of recovery for the drug respectively.

We can now look at some plots to see how this model plays out for various combinations of parameters:





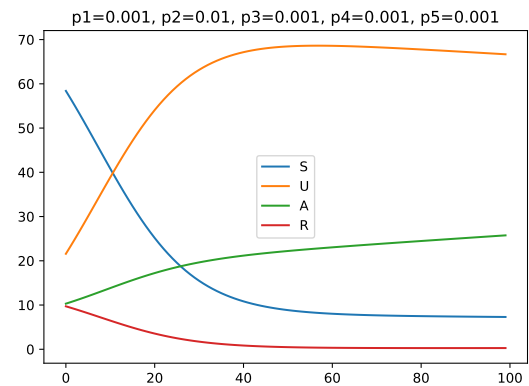
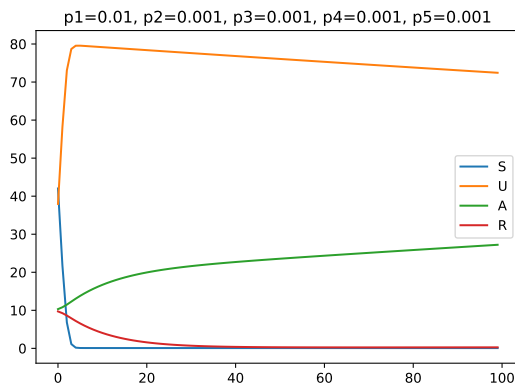
SUAR Model

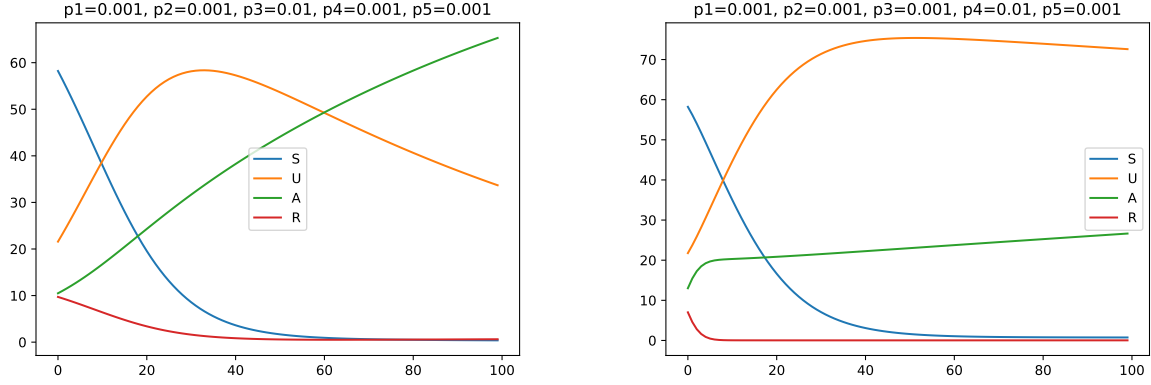
The SIR model fails to distinguish between different severities of addiction. By grouping anyone who is trying the drug once to people who are addicted in the same group we are failing to capture the behaviour of people trying the drug without becoming reliant on the drug. Because of this, we adapted the SIR model, renaming the infected group to addicted and adding in a new group for users of the drug who are not addicted, labeled as the ‘Using’ group. Suseptible people will now move into the Using group before moving into the addicted group. Since there are now two groups who are using the drug and interacting with the community, the I terms in the ODEs desribing the SIR now have to be replaced with $U + A$

The new model can now be described with the folowing equations:

$$\begin{aligned}\frac{dS}{dt} &= -\pi_1 * S * (A + U) + \pi_2 * U \\ \frac{dU}{dt} &= \pi_1 * S * (A + U) - \pi_2 * U - \pi_3 * U \\ \frac{dA}{dt} &= \pi_3 * U + \pi_4 * (A + U) * R - \pi_5 * A \\ \frac{dR}{dt} &= \pi_5 * A - \pi_4 * (A + U) * R\end{aligned}$$

With parameters $\pi_1, \pi_2, \pi_3, \pi_4$, and π_5 being the ‘infectivness’ of the drug for suseptible people, the rate that people stop trying the drug, the rate that people bome addicted, the ‘infectiveess’ for recovered people and the recovery rate for addicted people respectively.





SUAR Model with age

Looking at the stationary points of the previous two models (see appendix for derivation), the only long term behaviour is to either kill off the drug and have no users or addicted people or have everyone in the addicted or recovered group. However, this behaviour is not consistent with what we see in the real world, where we have a consistent number of people using and addicted to the drug, and a consistently high number of people who are susceptible to the drug. To fix we add age groups into the model, grouping them as children, teens, young adults, adults and seniors. Each of these groups have a distinct rate of death and births are proportional to the number of young adults and adults. Separating these groups also allows us to set different parameters according to each age group, capturing more nuanced behaviours such as more experimentation in young people.

Another feature we wanted to tackle is how people recover. In the same way that people start using drugs based on interactions with others who are using the drug, we expect people to recover based on interactions with people who aren't using the drug. This can be represented by support groups, concerned parents or friends. Furthermore, we can expect people in one age group to be influenced by different amounts from each age group. We can store this information with a matrix I , where I_{ij} is the magnitude of the influence that group j has on group i . We can now go in to further simplify this by defining two new terms, the positive and negative influence on a age group i , to be:

$$P_i = \sum_k I_{ik} * (S_k + R_k)$$

$$N_i = \sum_k I_{ik} * (U_k + A_k)$$

The model now has 20 compartments with both sideways movement within age groups and downwards movement as people age.

Looking across age group 'i', the model can be described as:

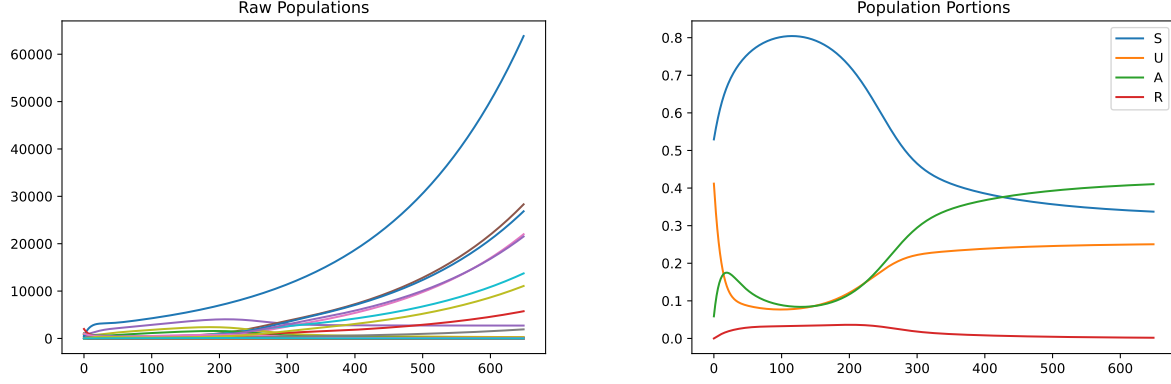
$$\frac{dS_i}{dt} = aS_{i-1} + aU_{i-1}\pi_{i-1,2}P_{i-1} - S_i\pi_{i,1}N_i + (1-a)U_i\pi_{i,2}P_i - aS_i - d_{i,s}S_i + (b \sum_{k=2}^3 S_k + U_k + A_k + R_k | i=0)$$

$$\frac{dU_i}{dt} = aS_{i-1}\pi_{i-1,1}N_{i-1} + aU_{i-1} - U_i * \pi_{i,3} - aU_i - U_i\pi_{i,2}P_i + (1-a)S_i\pi_{i,1}N_i - d_{i,U}U_i$$

$$\frac{dA_i}{dt} = aU_{i-1}\pi_{i-1,3} + aA_{i-1} + aR_{i-1}\pi_{i-1,4}N_{i-1} - A_i\pi_{i,5}P_i + (1-a)R_i\pi_{i,4}N_i - aA_i + (1-a)U_i\pi_{i,3} - d_{i,A}A_i$$

$$\frac{dR_i}{dt} = aA_{i-1}\pi_{i-1,5}P_{i-1} + aR_{i-1} - aR_i - R_i\pi_{i,4}N_i + (1-a)A_i\pi_{i,5}P_i - d_{i,R}R_i$$

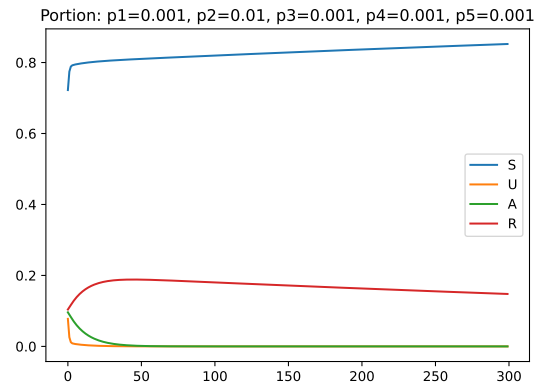
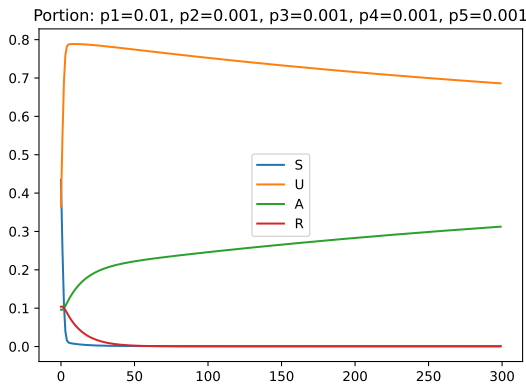
With the parameters being sorted in matrices π , $\{d\}$ and $\{i\}$, containing row-wise versions of the parameters for the SUAR model, the death rate for each compartment, and the influence each age group has on the other, and scalars a and b , which define the aging rate and the birth rate of the population.

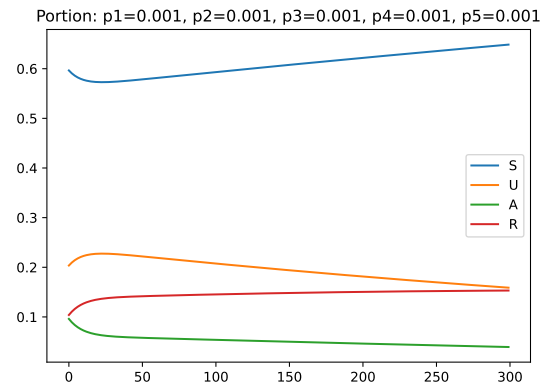
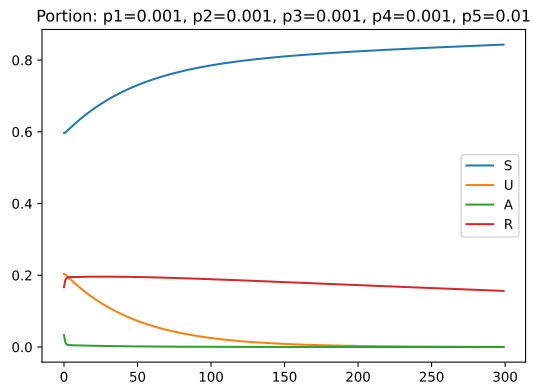
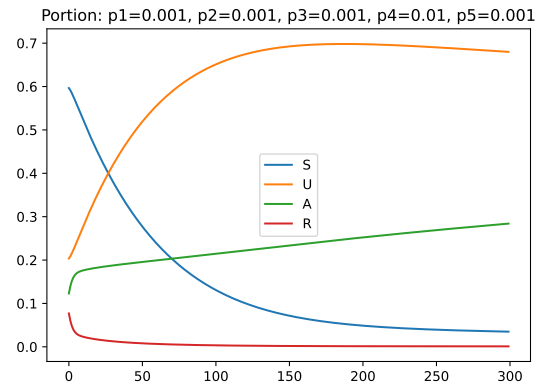
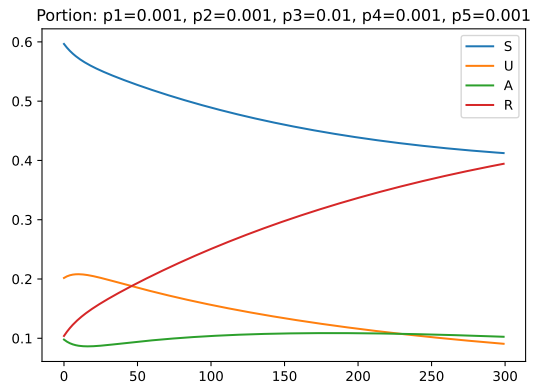


Simplified SUAR Model with age

After building the SUAR model with different age groups, we toyed around with various combinations of parameters to see how the model reacted. However, we found that with 20 compartments and 73 parameters, it is really hard to go through all the combinations to find meaningful results from the model. We could see that there were stationary points for the population portions, but finding them analytically was near impossible. Because of this, we sought to find a way to simplify the model to produce results that are easier to understand. We did this by removing the different age groups all together, while keeping the equations describing the model the same. By removing the age groups, the equations dramatically simplify into this form;

$$\begin{aligned}\frac{dS}{dt} &= -S\pi_1N + U\pi_2P - d_S S + b(S + U + A + R) \\ \frac{dU}{dt} &= S\pi_1N - U\pi_2P - U\pi_3 - d_U U \\ \frac{dA}{dt} &= U\pi_3 - A\pi_4P + R\pi_5N - d_A A \\ \frac{dR}{dt} &= A\pi_4P - R\pi_5N - d_R R\end{aligned}$$





Results

// Stationay distributions of final model -> jacobian plot

Discussion

// Compare our model // Inspirations // Limitaion // Future stuff

Concnclution

// Conclution

Appendix

Appendix 1 : Stationary points of SIR model

$$\begin{aligned}0 &= -\pi_1 * S * I \\0 &= \pi_1 * S * I + \pi_2 * R * I - \pi_3 * I \\0 &= -\pi_2 * R * I + \pi_3 * I\end{aligned}$$

Therefore either π_1 , S or I must be 0

If $I = 0$:

$$\begin{aligned}0 &= -\pi_1 * S * 0 \\0 &= \pi_1 * S * 0 + \pi_2 * R * 0 - \pi_3 * 0 \\0 &= -\pi_2 * R * 0 + \pi_3 * 0\end{aligned}$$

Therefore any point $[S, 0, R]$ is stationary.

If $\pi_1 = 0$ or $S = 0$:

$$\begin{aligned}0 &= 0 * I \\0 &= 0 * I + \pi_2 * R * I - \pi_3 * I \\0 &= -\pi_2 * R * I + \pi_3 * I\end{aligned}$$

Therefore:

$$\begin{aligned}\pi_2 * R * I - \pi_3 * I &= -\pi_2 * R * I + \pi_3 * I \\ \pi_2 * R * I &= \pi_3 * I \\ R : I &= \pi_3 : \pi_2\end{aligned}$$

Therefore any point $[0, \frac{P_{total} * \pi_2}{\pi_2 + \pi_3}, \frac{P_{total} * \pi_3}{\pi_2 + \pi_3}]$

Appendix 2 : Stationary points of SUAR

$$\begin{aligned}0 &= -\pi_1 * S * (A + U) + \pi_2 * U \\0 &= \pi_1 * S * (A + U) - \pi_2 * U - \pi_3 * U \\0 &= \pi_3 * U + \pi_4 * (A + U) * R - \pi_5 * A \\0 &= \pi_5 * A - \pi_4 * (A + U) * R\end{aligned}$$

By combining the first two and last two equations, one can see that if the parameters are non 0, $U = 0$.
Therefore:

$$\begin{aligned}0 &= \pi_1 * S * A \\0 &= \pi_4 * A * R - \pi_5 * A \\0 &= \pi_5 * A - \pi_4 * A * R\end{aligned}$$

Therefore $S = 0$ or $A = 0$ In the case $A = 0$, any point $[S, 0, 0, R]$ will be stationary

In the case $S = 0$

$$\pi_4 * A * R = \pi_5 * A$$

$$R : A = \pi_5 : \pi_4$$

Therefore any point $[0, 0, \frac{P_{total} * \pi_4}{\pi_4 + \pi_5}, \frac{P_{total} * \pi_5}{\pi_4 + \pi_5}]$