

## Heroin Model

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### Background

The misuse of opioids, a drug class including prescription pain relievers and the illegal drug heroin, is rampant in today's society. The opioid crisis was declared a public health emergency in October 2017 by the United States Department of Health and Human Sciences [15]. In 2016, there were 11.8 million opioid misusers 12 years of age or older with 948,000 of these being heroin users [10]. In addition, there was an estimated 13,219 heroin deaths in 2016, a more than six-fold increase from the year 2002 [7]. This, in part, is due to the recent trend of lacing heroin with fentanyl, a surgical-grade opioid that is up to fifty times more potent than heroin alone, and therefore, users are unaware of the purity of the heroin they obtain [9,12].

Prescription pain relievers include oxycodone, hydrocodone, and morphine products, among others [18]. This drug class is misused for a variety of reasons with the most prominent being to relieve physical pain, to feel good or get high and to relax or relieve tension. Individuals that misuse prescription opioids mostly obtain them from friends/relatives or from a healthcare provider, with misuse defined as taking the prescription at a higher dose or more frequently than prescribed, or taking someone else's medication [10]. **Check on how this data was obtained/the survey questions asked, etc.** Opioids are of no shortage in society today and since opioid addiction is driven largely by legal prescription medication availability, this makes a significant proportion of society susceptible to misuse and addiction to opioids. This leads some individuals to start heroin, and in fact, 80% of heroin users used prescription opioids before their heroin use [32].

In the mid-1990's, the American Pain Society aggressively pushed the idea of pain as the fifth vital sign as they believed pain was being undertreated in doctor offices and hospitals. **Check on sources, if the APS was supported/funded by drug companies.** In 2000, the Joint Commission required physicians to accept and respect the self-reporting of pain by patients. By the early 2000s, drug manufacturers funded publications and physicians to support opioid use for pain control [31]. **Check if this the proof part is true: This is one factor in the near tripling of prescriptions that pharmacies distributed from 1991 to 2011, as doctors could use opioids as a proof of treating pain [14].** This was in part due to a number of new opioids that were approved by the FDA for use, such as OxyContin, Actiq, Fentora, and Onsolis, among others, in addition to other unapproved products [33].

Moreover, there is a higher availability of heroin in recent years at a lower cost than alternative opioids [14]. In the 1960's, heroin users were composed mainly of young, non-white men in urban areas with their initial opioid being heroin, but in present day, this trend has shifted to older, white, rural and suburban men and women with their initial opioid being a prescription [14].

To address this apparent problem in today's society, we wish to investigate the dynamics behind the opioid and heroin epidemic and identify important conditions relating to the reduction of opioid and heroin addicted individuals. To do that, we have formulated a population level system of ordinary differential equations model consisting of classes of individuals taking prescription opioids, addicted to opioids, using heroin and recovering from opioid addiction, including heroin, and analyzed it. Our overall goal is to explore management strategies for optimally treating pain with prescriptions while reducing opioid addiction and heroin use.

The motivation for this model came from a previous model focusing on opioid addicts through prescriptions or via the black market [1]. The main results included the following: include results from most recent update of submitted paper but for now: - In order to have an addiction-free equilibrium, both addictions that come from prescriptions and addictions from accessibility to excess drugs must be eliminated ( $\gamma = 0 = \xi$ ). - Near addiction-free state: prevention of prescription opioid users becoming addicted is more important than reducing prescriptions getting into the hands of non-prescribed users to combat the epidemic. - More realistically, outside of addiction-free state: reducing prescription-user addictions, decreasing prescriptions dispensed and increasing entry into treatment are most important to reduce number of addicted.

The purpose of formulating a separate model is to be able to understand the more complicated dynamics that arise among opioid addiction with the addition of heroin use. Incorporating heroin is important since it seems to play a role in the process of opioid addiction, transition to heroin and recovery process, and therefore, it's inclusion provides a more accurate overall picture of the epidemic. In addition, since heroin users overdose almost as much as prescription opioid users and in recent years, a sharp increase of heroin overdoses has occurred, it is crucial to take a more detailed look at this increasing problem [25].

ADD in more info about other heroin models (Markov chain, compartmental, etc.)

## Model Formulation

Our model consists of five subgroups of the (national) population, in which they are all proportions of the entire population:

1. Susceptibles ( $S$ ): This portion of the population consists of individuals who are not taking prescription opioids of any kind, nor are using heroin.
2. Prescription opioid users ( $P$ ): This class of individuals consists of individuals who are prescribed opioids by a health care provider and take the opioids at a level that is not considered addicted. We note that individuals in this class could be misusing opioids, but not at the level of addiction.
3. Opioid addicts ( $A$ ): This group of individuals are addicted to opioids or are dependent on opioids. Although the opioid class of drug includes heroin, here we will take opioids to mean non-heroin.

4. Heroin users ( $H$ ): This class is composed of individuals who use heroin, which is implicitly understood to be addictive. We note that individuals in this class could be using other drugs or are addiction to opioids in addition to heroin, but they are at least using heroin.

5. Individuals in treatment/rehabilitation ( $R$ ): This class consists of individuals undergoing treatment for their addiction to opioids and/or heroin.

We assume that our these five compartments sum to one, so that the total death rate is equal to the incoming rate for the susceptible class. Although terminology is not always clearly defined in the literature, such as opioid misuse, abuse, dependency, addiction and use disorder, we aim to focus on addiction only, where addiction to opioids is defined as having a pattern of continued use with the potential for harm [27]. We take opioid use disorder to fall in this categorization of addiction, due to the definition presented in [34] which includes sustained use regardless of interference with life obligations. We also recognize there is difficulty in modeling to the variability of the purity of heroin, as overdose deaths could come from the opioid fentanyl rather than the actual heroin itself, and in fact, one in five overdose deaths have multiple drugs present so it is difficult to know the actual cause of death [25]. We try to take into account this complexity as best as possible with the information in literature.

We denote the initial conditions as  $S(0) = S_0$ ,  $P(0) = P_0$ ,  $A(0) = A_0$ ,  $H(0) = H_0$ , and  $R(0) = R_0$ , and we assume all of these values are positive. From data, the initial values for each of the classes are the following proportions of the entire population, and therefore are dimension-less: [EDIT for local data in Knox County/Tennessee](#).

Here, we present our ordinary differential equation model:

$$\frac{dS}{dt} = -\alpha S - \beta(1-\xi)SA - \beta\xi SP - \theta_1 SH + \epsilon P + \delta R + \mu(P+R) + (\mu + \mu_A)A + (\mu + \mu_H)H \quad (1)$$

$$\frac{dP}{dt} = \alpha S - \epsilon P - \gamma P - \theta_2 PH - \mu P \quad (2)$$

$$\frac{dA}{dt} = \gamma P + \sigma_A R + \beta(1-\xi)SA + \beta\xi SP - \zeta A - \theta_3 AH - (\mu + \mu_A)A \quad (3)$$

$$\frac{dH}{dt} = \theta_1 SH + \theta_2 PH + \theta_3 AH + \sigma_H R - \nu H - (\mu + \mu_H)H \quad (4)$$

$$\frac{dR}{dt} = \zeta A + \nu H - \delta R - \sigma_A R - \sigma_H R - \mu R \quad (5).$$

The parameters involved in this model represent transition rates from one class to another and are per capita yearly rates; specifically:

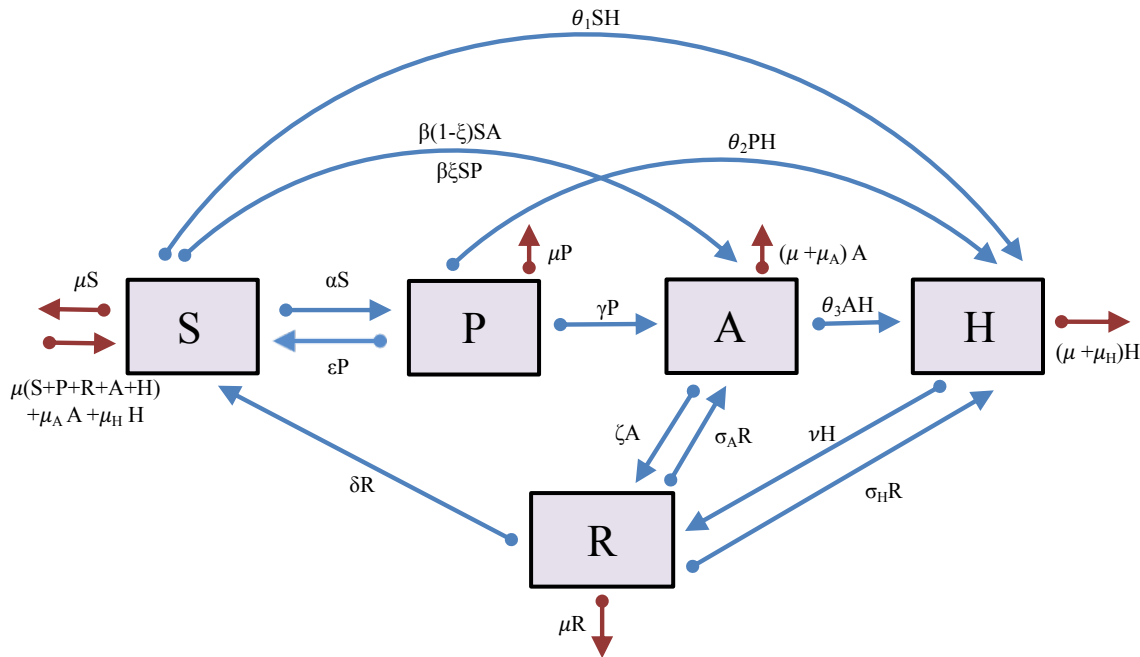
- $\alpha S$ : rate at which susceptible individuals are prescribed opioids (1/year)
- $\beta$ : rate at which susceptible individuals become addicted to opioids by some means other than by prescription (1/year)

- $\beta(1 - \xi)SA$ : rate at which susceptible individuals become addicted to opioids by black market drugs or interaction with other addicts (1/year, dimension-less)
- $\beta\xi SP$ : rate at which susceptible population obtains extra prescription opioids and becomes addicted (1/year, dimension-less)
- $\theta_1 SH$ : rate at which susceptible population becomes addicted to heroin by black market availability or interaction with other heroin users (1/year)
- $\epsilon P$ : rate at which individuals return to the susceptible class after being prescribed opioids and did not develop an addiction (1/year)
- $\delta R$ : rate at which both opioid and heroin addicts enter back into the susceptible class after successfully finishing treatment (note: successfully finishing treatment does not protect individuals from future addiction, as there is no recovery) (1/year)
- $\mu S, \mu P, \mu A, \mu H, \mu R$ : natural death rates (1/year)
- $\mu_A A$ : enhanced death rate for opioid addicts; overdose rate which results in death (1/year)
- $\mu_H H$ : enhanced death rate for heroin addicts; overdose rate which results in death (1/year)
- $\gamma P$ : rate at which prescribed opioid users become addicted to opioids (1/year)
- $\theta_2 PH$ : rate at which prescribed opioid users become addicted to heroin (1/year)
- $\sigma_A R$ : rate at which individuals transition from treatment into the opioid addicted class (note: not considered relapse because do not know where individuals came from originally) (1/year)
- $\zeta A$ : rate at which addicted opioid users enter treatment/rehabilitation (1/year)
- $\theta_3 AH$ : rate at which the opioid addicted population becomes addicted to heroin (1/year)
- $\sigma_H H$ : rate at which individuals transition from treatment into the heroin addicted class (note: not considered relapse because do not know where individuals came from originally) (1/year)
- $\nu H$ : rate at which heroin users enter treatment/rehabilitation

Since susceptible individuals die and give birth at the same rate,  $\mu S$ , those terms do not appear in the differential equation for susceptibles. Also, we note that although sellers are not directly involved in the model nor have to be addicts themselves, addicted individuals act as a proxy for sellers of the drug and general availability of the drug, whether it is opioids or heroin. This is because the higher number of addicted individuals there are, the more sellers there needs to be which results in an increased exposure to the drug and therefore, increased chance of addiction. We do not include an interaction term between addicted and those in recovery to go from the recovery to addicted class

because we assume that individuals in recovery already know the feeling of addiction and do not need the influence of addicted individuals or general availability of the drug to fall into addiction themselves. Moreover, we omitted an interaction term between prescribed users and addicted individuals moving into the addicted class, under the assumption that since prescribed users have access to their own prescriptions to get addicted, they do not need an illicit source or interaction with addicted individuals to become addicted themselves. We note that there are treatments available for opioid and heroin use, which involves medication such as Buprenorphine, Methadone and Naltrexone, in combination with counseling and behavioral therapies [26]. Although treatments for addictions are an enormous problem in and of themselves since there is limited success, we are incorporating treatments at a shallow level as a part of our model.

This compartmental model can be represented by the following flow diagram with each arrow representing either the transition rate between one class to another or birth/death:



## Parameter Values

EDIT for local data in Knox County/Tennessee: May use ordinary least squares or bayesian inference since we are hoping to have data from experts—think about after NC State Parameter Estimation workshop

## Proof of existence of solution to ODE system

Will show existence of solution to my ODE system.

## Proof of non-negative initial conditions guaranteeing non-negative solutions

We will now show that starting with non-negative initial conditions, solutions for the system (1)-(5) are guaranteed to be non-negative for all time, using the following theorem which is an adaptation of Theorem 2.1 in [17]:

**Theorem 1** ([17]: adapted version of Theorem 2.1, Chapter 5) *Assume that when  $\phi \in \mathbb{R}^n$  satisfies  $\phi \geq 0$  with  $\phi_i(0) = 0$  for some  $i$ , then  $f_i(\phi) \geq 0$ . Then, if  $\phi \in \mathbb{R}^n$  satisfies  $\phi \geq 0$ , the solution of  $x'(t) = f(x(t))$ ,  $x(t_0) = \phi$  where  $f : \mathbb{C} \rightarrow \mathbb{R}^n$  continuous,  $\mathbb{C} \subset \mathbb{R}^n$  open and  $\phi \in \mathbb{R}^n$ , satisfies  $x(t) \geq 0$  for all  $t \geq t_0$ .*

**Proof** Consider the vector of initial conditions in  $\mathbb{R}^5$ ,  $\phi = (S_0, P_0, A_0, H_0, R_0)$ . Since each vector component represents a proportion of a population,  $\phi \geq 0$ .

1) Assume  $\phi = (0, P_0, A_0, H_0, R_0) \geq 0$ . Then,

$$f_1(\phi) = \epsilon P_0 + \delta R_0 + \mu(P_0 + R_0) + (\mu + \mu_A)A_0 + (\mu + \mu_H)H_0 \geq 0,$$

since all parameters and variables are non-negative.

2) Assume  $\phi = (S_0, 0, A_0, H_0, R_0) \geq 0$ . Then,

$$f_2(\phi) = \alpha S_0 \geq 0,$$

since  $\alpha$  and  $S_0$  are non-negative.

3) Assume  $\phi = (S_0, P_0, 0, H_0, R_0) \geq 0$ . Then,

$$f_3(\phi) = \gamma P_0 + \sigma_A R_0 + \beta \xi S_0 P_0 \geq 0,$$

since all parameters and variables are non-negative.

4) Assume  $\phi = (S_0, P_0, A_0, 0, R_0) \geq 0$ . Then,

$$f_4(\phi) = \sigma_H R_0 \geq 0,$$

since  $\sigma_H$  and  $R_0$  are non-negative.

5) Assume  $\phi = (S_0, P_0, A_0, H_0, 0) \geq 0$ . Finally,

$$f_5(\phi) = \zeta A_0 + \nu H_0 \geq 0,$$

since all parameters and variables are non-negative.

### Addiction-Free Equilibrium

To find the addiction-free equilibrium, we set equations (1)-(5) equal to zero and require that  $A = H = R = 0$  since these three compartments consist of individuals who are addiction to opioids or heroin. We are left with the system:

$$0 = -\alpha S^* - \beta \xi S^* P^* + \epsilon P^* + \mu P^*$$

$$0 = \alpha S^* - \epsilon P^* - \gamma P^* - \mu P^*$$

$$0 = \gamma P^* + \beta \xi S^* P^*.$$

If  $P^* = 0$ , then the only solution is  $S^* = P^* = H^* = R^* = 0$ , but  $S^* = P^* = H^* = R^* = 1$  by assumption, so this is not possible. Thus, we will assume  $P^* \neq 0$ . This forces  $\gamma + \beta \xi S^* = 0$  and since all of our parameters and variables are non-negative, then it must be  $\gamma = 0$  and either  $\beta = 0$  or  $\xi = 0$ . We have that  $\gamma = 0$  means that individuals who are prescribed opioids cannot become addicted to opioids. In addition,  $\xi = 0$  means that only black market opioids are available and there are no excess prescription drugs available and  $\beta = 0$  means that susceptibles are unable to become addicted to opioids at all. We claim that  $\beta = 0$  is less realistic than the former since eliminating two paths to addiction from the susceptible class is less realistic than just eliminating one path. Under the assumption that  $\gamma = 0 = \xi$  to ensure the existence of our addiction-free equilibrium and that  $1 = S + P + A + H + R$ , we calculate the addiction-free equilibrium to be:

$$S^* = \frac{\epsilon + \mu}{\alpha + \epsilon + \mu}$$

$$P^* = \frac{\alpha}{\alpha + \epsilon + \mu}$$

$$A^* = 0$$

$$H^* = 0$$

$$R^* = 0$$

Note that enforcing  $P^* \neq 0$  implies that  $\alpha \neq 0$ , as well, which means that individuals are still able to be prescribed opioids.

### Basic Reproduction Number, $\mathcal{R}_0$

The basic reproduction number, denoted  $\mathcal{R}_0$ , is a term used in epidemiological models that gives the expected number of secondary disease cases that result from the introduction of a disease to a susceptible population. The value of  $\mathcal{R}_0$  represents how successful the spread of the disease is expected to be; if  $\mathcal{R}_0 < 1$ , then the disease is expected to die out and the disease-free equilibrium will be locally stable; conversely, if  $\mathcal{R}_0 > 1$ , then the

disease is expected to spread and the disease-free equilibrium will be unstable [11].

This idea may be applied to the context of our model since for the addiction-free equilibrium to occur,  $\gamma = 0$  and  $\xi = 0$ , so that individuals can become addicted only with interactions with opioid addicted individuals or heroin users so takes the form of an infectious disease. Thus, in this case in which the prescription part is essentially taken out of the model so that it is a black-market-only driven model,  $\mathcal{R}_0$  can be calculated. At the addiction-free equilibrium, the population is completely susceptible, as needed for  $\mathcal{R}_0$  to be an accurate measurement of addiction potential. We note that our goal in this section is to explore the structure of the model and not obtain a result for the full model. The disease compartments are those that contain infected individuals, or in our case, addicts [11]. Our model incorporates three addiction compartments, A, H, and R, since these all consist of opioid and/or heroin addicted individuals. We will utilize the Next Generation Matrix Method in order to calculate  $\mathcal{R}_0$ .

For the purposes of calculating  $\mathcal{R}_0$ , we will assume  $\gamma = 0$  and  $\xi = 0$  (thus,  $\beta \neq 0$ ) in order to ensure the existence of the addiction-free equilibrium. This results in the following system:

$$\begin{aligned}\frac{dS}{dt} &= -\alpha S - \beta SA - \theta_1 SH + \epsilon P + \delta R + \mu(P + R) + (\mu + \mu_A)A + (\mu + \mu_H)H \\ \frac{dP}{dt} &= \alpha S - \epsilon P - \theta_2 PH - \mu P \\ \frac{dA}{dt} &= \sigma_A R + \beta SA - \zeta A - \theta_3 AH - (\mu + \mu_A)A \\ \frac{dH}{dt} &= \theta_1 SH + \theta_2 PH + \theta_3 AH + \sigma_H R - \nu H - (\mu + \mu_H)H \\ \frac{dR}{dt} &= \zeta A + \nu H - \delta R - \sigma_A R - \sigma_H R - \mu R.\end{aligned}$$

In general, the differential equations of the  $n$  disease compartments,  $x'_i$ , may be written as:

$$x'_i = \mathcal{F}_i(x, y) - \mathcal{V}_i(x, y), \quad i = 1, \dots, n$$

where  $x$  is a vector with the disease compartments as components,  $y$  a vector with the non-disease compartments as components,  $\mathcal{F}_i$  represents the rate that secondary infections contribute to disease compartment  $i$  and  $\mathcal{V}_i$  represents the rate of transitions, i.e. rate at which the disease compartment  $i$  is decreased by means of death, recovery and progression of the disease for  $i = 1, 2, 3$  [11].

Thus, for our three addicted compartments, we may write:

$$\frac{dA}{dt} = \mathcal{F}_1(x, y) - \mathcal{V}_1(x, y)$$



$$\begin{aligned}\frac{dH}{dt} &= \mathcal{F}_2(x, y) - \mathcal{V}_2(x, y) \\ \frac{dR}{dt} &= \mathcal{F}_3(x, y) - \mathcal{V}_3(x, y),\end{aligned}$$

where  $x = [A \ H \ R]^T$  and  $y = [S \ P]^T$ .

Thus, under the assumption that A, H and R are the addicted compartments and abiding by the parameter restrictions stated above, the assumptions of the Next Generation Method are satisfied for matrices  $\mathcal{F}$  and  $\mathcal{V}$  formulated here:

$$\mathcal{F} = \begin{pmatrix} \beta SA \\ \theta_1 SH + \theta_2 PH \\ 0 \end{pmatrix}$$

$$\mathcal{V} = \begin{pmatrix} -\sigma_A R + \zeta A + \theta_3 AH + (\mu + \mu_A)A \\ -\theta_3 AH - \sigma_H R + \nu H + (\mu + \mu_H)H \\ -\zeta A - \nu H + \delta R + \sigma_A R + \sigma_H R + \mu R \end{pmatrix}.$$

These assumptions include the following:

- $\mathcal{F}_i(0, y) = 0$  and  $\mathcal{V}_i(0, y) = 0, \forall y \geq 0, i = 1, 2, 3$ , which ensures that new addictions arise only from interacting with those currently addicted, and there is no immigration into the addicted compartments; these guarantee that the population free of addiction remains that way.
- $\mathcal{F}_i(x, y) \geq 0, \forall x, y \geq 0, i = 1, 2, 3$  since these represent new addictions.
- $\mathcal{V}_i(x, y) \leq 0$  whenever  $x_i = 0, i = 1, 2, 3$  since there must be inflow only when the respective compartment is empty.
- $\sum_{n=1}^3 \mathcal{V}_i(x, y) \geq 0, \forall x, y \geq 0$  since this represents the total outflow from all addicted compartments.
- The addiction-free system,  $\left. \frac{dS}{dt} \right|_{A,H,R=0}$  and  $\left. \frac{dP}{dt} \right|_{A,H,R=0}$  has a unique equilibrium,  $(\frac{\epsilon+\mu}{\alpha+\epsilon+\mu}, \frac{\alpha}{\alpha+\epsilon+\mu})$  that is asymptotically stable.

Taking  $F = \frac{\partial \mathcal{F}_i}{\partial x_j}(0, y_0)$  and  $V = \frac{\partial \mathcal{V}_i}{\partial x_j}(0, y_0), i, j = 1, 2, 3$ , where  $(0, y_0) = (\frac{\epsilon+\mu}{\alpha+\epsilon+\mu}, \frac{\alpha}{\alpha+\epsilon+\mu}, 0, 0, 0)$  is the addiction-free equilibrium, we calculated the following [11]:

$$F = \begin{pmatrix} \beta S^* & 0 & 0 \\ 0 & \theta_1 S^* + \theta_2 P^* & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$V = \begin{pmatrix} \zeta + \mu + \mu_A & 0 & -\sigma_A \\ 0 & \nu + \mu + \mu_H & -\sigma_H \\ -\zeta & -\nu & \delta + \sigma_A + \sigma_H + \mu \end{pmatrix}.$$

The eigenvalues of  $FV^{-1}$  are calculated to be:

$$\sigma(FV^{-1}) = \left\{0, \frac{(r+s) - \sqrt{(r-s)^2 + 4\beta S^* z \sigma_A \zeta \sigma_H \nu}}{2\det(V)}, \frac{(r+s) + \sqrt{(r-s)^2 + 4\beta S^* z \sigma_A \zeta \sigma_H \nu}}{2\det(V)}\right\}$$

where  $a = \zeta + \mu + \mu_A$ ,  $b = \nu + \mu + \mu_H$ ,  $c = \delta + \sigma_A + \sigma_H + \mu$ ,  $z = \theta_1 S^* + \theta_2 P^*$ ,  $r = \beta S^*(bc - \sigma_H \nu)$ ,  $s = z(ac - \sigma_A \zeta)$ , and  $\det(V) = a(bc - \sigma_H \nu) - \sigma_A \zeta b$ .

$\mathcal{R}_0$  may then be determined as the spectral radius of  $FV^{-1}$ , in which the  $(i, j)$  entry is the expected number of secondary additions in compartment  $i$  produced by individuals initially in compartment  $j$  :

$$\mathcal{R}_0 = \frac{(r+s) + \sqrt{(r-s)^2 + 4\beta S^* z \sigma_A \zeta \sigma_H \nu}}{2\det(V)}.$$

This value is the spectral radius since it is the largest eigenvalue greater than 0. We can see this by first noting that the radicand  $(r-s)^2 + 4\beta S^* z \sigma_A \zeta \sigma_H \nu$  is positive, since all parameters are positive. In addition,  $r$  is positive because in the multiplication of  $bc = \delta\mu + \delta\nu + \delta\mu_H + \mu^2 + \mu\nu + \mu\mu_H + \mu\sigma_A + \mu\sigma_H + \nu\sigma_A + \sigma_H\nu + \mu_H\sigma_A + \mu_H\sigma_H$ , all terms are positive and the highlighted term cancels with  $-\sigma_H\nu$ , leaving only a positive sum. Multiplying by  $\beta S^*$ , the resulting value  $r$  is positive. Similarly,  $s$  is positive because in the multiplication of  $ac = \delta\mu_A + \delta\mu + \delta\zeta + \mu_A\mu + \mu_A\sigma_A + \mu_A\sigma_H + \mu^2 + \mu\zeta + \mu\sigma_A + \mu\sigma_H + \sigma_A\zeta + \zeta\sigma_H$ , all terms are positive and the highlighted term cancels with  $-\sigma_A\zeta$ , again resulting in a sum of only positive terms. Multiplied against positive value  $z$ , this results in  $s$  being positive. Finally, the expansion of  $\det(V)$  is positive. This is because  $bc$  is a positive sum containing a term that cancels with  $-\sigma_H\nu$  as described above and  $ac$  shown above contains the term  $\sigma_A\zeta$  which multiplied with  $b$  cancels with  $-\sigma_A\zeta b$ , thus resulting in a positive determinant.

Increasing  $\beta$  (the rate of addiction via non-prescription means),  $\theta_1$  (the rate of heroin addiction of the susceptible population), and  $\theta_2$  (the rate of heroin addiction of the prescribed population), all increase  $\mathcal{R}_0$  since these parameters only appear in the numerator of  $\mathcal{R}_0$ , specifically within the  $r$ ,  $z$ , and  $s$  terms.

## Sensitivity Analysis

Argue why using Sobol versus others, such as Latin Hypercube and read more about how it works.

In order to explore the sensitivity of each of the population classes to each of the parameters, we implemented Sobol sensitivity analysis in which the Saltelli sampler is utilized to

generate  $N(2D+2)$  parameter samples where  $N=\text{FILL IN}$  is the number of sample points and  $D=16$ , the number of parameters in the model for a total of  $\text{FILL IN}$  samples [16]. Saltelli are the statistically good points chosen in the parameter space to test and records  $S$ ,  $P$ ,  $A$ ,  $H$  and  $R$  at the final time ( $t=10$  years) for the parameter choices. Samples  $N$  points from the problem and returns a matrix of parameter values.

Both  $\mu_A$  and  $\mu_H$  were restricted to the domain  $[0, 0.1]$  with the remaining parameters ranging on the entire domain  $[0, 1]$ . For first-order indices, only one parameter is varied and the rest are held constant; for second-order indices, two parameters are varied and the rest are held constant; and for total-order indices, every combination of parameters is varied, from a single parameter varying alone to all higher-order interactions between parameters. The length of the colored bars, with each color representing a specific class, measures the contribution of a certain parameter to the variance of each of the classes in the model. The longer the colored bar, the higher the effect the parameter has on that class of individuals. We are able to retrieve the confidence intervals on the sensitivity of each of the populations to a respective change in the parameter(s) for first order changes, second order changes and total order changes.

## Future Work

- Obtain local data from Knox County/East Tennessee, can calculate  $\mathcal{R}_0$  from there (would expect  $< 1$  b/c prescriptions seem to be a huge driving force behind addiction to opioids and individuals switching to heroin use. Contact with Davide Stern (UT hospital), Paul Erwin, Agricola Odoi, Kelly Cooper
- Perform sensitivity analysis to determine the sensitivity of each of the classes to the parameters (i.e. the contribution of each of the parameters to the sizes of the classes). Plan to use Sobol sensitivity analysis in which the saltelli sampler is utilized to generate (statistically good) parameter samples. Will only perform on the full model, not the AFE b/c that would require  $t$  very large and we want to look at shorter term, such as  $t = 10$  years. Will perform first, second and total indices.
- Fit parameters to data that are difficult to find in literature.
- Explore management strategies for how to best treat pain with prescriptions while reducing opioid addiction and heroin use.
- Break recovery class into two different classes for opioid addicts and heroin users.
- Look into gender, race, socioeconomic class or rural versus urban location to investigate differences