

# Modelling Outbreaks of Unhealthy Study Habits on Campus.

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GitHub link to code: <https://github.com/paszach/zombies-attack>

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## 1 Context

An unexpected outbreak has recently spread through Imperial, causing chaos during deadline season. Students with once healthy study habits are having their time management skills corrupted by contagious bad academic habits.

Remaining healthy students are being cautioned to stay away from individuals that appear to be stumbling around aimlessly after carrying out all-nighters. Disaster analysts have determined that the outbreak of students with bad habits, dubbed the 'Cooked,' for their seemingly fried brains, originated from the underneath the Huxley tunnels.

Although the outbreak started in the JMC department, it was quickly observed that the unhealthy habits that make students Cooked spread throughout group projects, leading to Cooked students passing on their unhealthy academic habits to students with healthy habits. Through multi-department modules, these habits started to spread across the entirety of campus. After these bad habits take hold, students lose the ability to keep on top of their work, resulting in bad grades, or worse—failing to the point of dropping out.

In an attempt to understand and hopefully mitigate the impact of this outbreak, I have been enlisted as the only remaining student with a educational license to Wolfram Mathematica. My aim is to utilise my modelling and simulation knowledge to model the current known characteristics of the outbreak in order to determine the fate of Imperial as we know it. I will assess different strategies to combat the spread of unhealthy habits, evaluating my model as I go.

If I get Cooked whilst writing this report, I've uploaded all my code to GitHub, in hope it will be found and used for good: <https://github.com/paszach/zombies-attack>

This report begins with constructing a basic model to represent the spread of the Cooked. Improvements will be made on the model to closer reflect reality, culminating in a strategy on how Imperial can prevent (or control) a disastrous outbreak.

From initial observations, an SIR model will be used as a starting point.

## 2 Basic SIR Model

The SIR model has been proven vital in understanding the approach used to tackle pandemics such as COVID-19. The basic SIR model is a compartmental system which has three states a member of the population can be in.

**Susceptible:** Those who have not yet been infected.

**Infected:** Those who spread disease by infecting members of the susceptible population.

Once infected, infected members stay infected and contagious for a period of time.

**Removed:** Those who were infected then leave the infectious period and enter the removed population. Members of the removed population are unable to become reinfected.

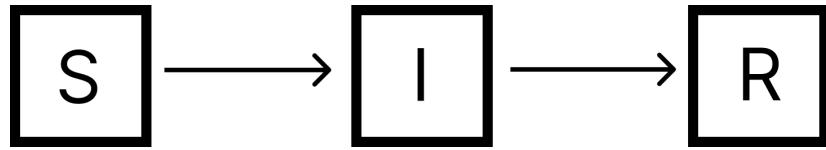


Figure 1: Diagram of basic SIR model.

The SIR model can also be adapted to model the outbreak of Cooked students, since the spread of unhealthy habits can be likened to an infection[1]. Bad study have been seen to spread through group projects to students with healthy habits. Once bad habits take hold, students are doomed to fail and be forced to drop out. This model will be the starting point of simulating the Imperial outbreak, and can be contextualised as an altered 'HCR' model;

**Healthy:** These students (members of the Healthy student population) have healthy study habits. They are susceptible to being influenced by the bad study habits of Cooked students, resulting in them joining the Cooked population.

**Cooked:** Cooked students can infect healthy students with their unhealthy study habits. Once bad habits have taken hold, students' grades decline, until they fail the course, and must drop out.

**Removed:** Once students have dropped out due to their failing grades, they cannot return to their studies (re-enter the Healthy or Cooked student population).

## 2.1 Assumptions

Several assumptions have been made in order to comply with the limitations of an HCR (SIR) model;

**Closed Population:** The total population of the student body is a constant,  $P = H + C + R$ .  
No new students are enrolling to Imperial, and none are leaving for reasons other than being forced to drop out.

**Homogenous Mixing:** Every individual student is equally likely to interact with every other student. There is no separation between courses, department, or undergrad/postgrad students.

**Identical Individuals:** Students are equally likely to be influenced by bad study habits, and equally 'infectious' once bad study habits have been contracted.

**Removal of Events:** There are no events which have influence over the spread-ability of bad habits, such as deadlines or graduation.

**Compartments are discrete:** Students move instantaneously between H, C and R states, as opposed to bad habits slowly setting in.

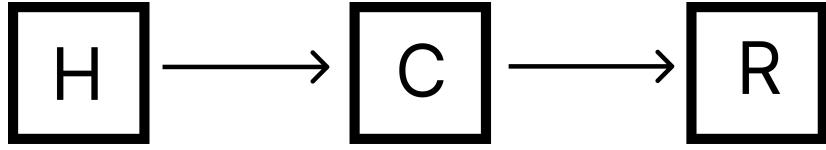


Figure 2: Diagram of basic HCR model.

## 2.2 Parameters

The parameters that drive the HCR model have been added to the diagram (Fig. 3).

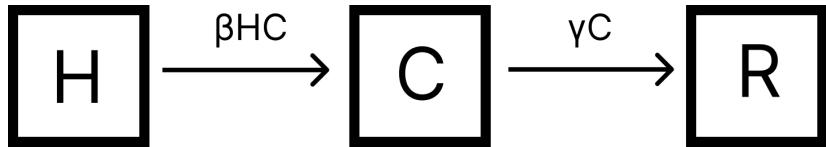


Figure 3: Diagram of basic HCR model with labelled transition rates.

$\beta_c$ : The transmission parameter  $\beta_c$  defines the rate at which a Cooked student transmits their bad habits to healthy students. If the number of healthy students is  $H$ , and  $P$  is the total population, the average rate at which one Cooked student spreads their bad habits is  $\beta_c \frac{H}{P}$  per unit time.

Therefore, the total number of new Cooked students per unit time can be expressed as a product of the per Cooked student spread rate  $\beta_c \frac{H}{P}$  and the total number of Cooked students,  $C$ :

$$(\beta_c) \left( \frac{H}{P} \right) C, P = 1 \therefore C' = \beta_c H C$$

$\gamma_c$ : This is the removal, or drop-out rate that at which the Cooked drop out of Imperial. Dropping out is independent of the healthy student population, i.e. not due to healthy students encouraging cooked students to drop out or any other factor driven by  $H$ .

In real life, Imperial drop out rates tend to be 1.55%[2]. A range of between 0 and 5% will be used to investigate the system.

## 2.3 Governing Equations

Now the parameters are defined, the basic set of equations that form the HCR model can be shown. Each equation shows a change in population,  $H$ ,  $C$  or  $R$ . Positive terms show that students are joining the population, and negative terms show students that are leaving the population.

$$\begin{aligned} H' &= -\beta_c H C \\ C' &= \beta_c H C - \gamma_c C \\ R' &= \gamma_c C \end{aligned}$$

## 2.4 Natural Units

Since the population has already been normalised ( $H + C + R = 1$ ),  $\beta_c$  is the only term in the governing equations that has units: ( $t^{-1}$ ). Therefore the equation can be non-dimensionalised by using the following equation.

$$t = k\hat{t}$$

Where  $k$  has units of time,  $t$ . From here, the relationship between  $\frac{d}{dt}$  and  $\frac{d}{d\hat{t}}$  can be determined;

$$\frac{d}{dt} = \frac{d\hat{t}}{dt} \cdot \frac{d}{d\hat{t}} = \frac{1}{k} \cdot \frac{d}{d\hat{t}}$$

Now, applying this to the original first equation of the healthy population;

$$H' = \frac{dH}{dt} = \frac{d}{dt}(H) = \frac{1}{k} \cdot \frac{dH}{d\hat{t}} = -\beta_c HC \therefore \frac{dH}{d\hat{t}} = -k\beta_c HC$$

From this point onward, primes represent derivatives with respect to the dimensionless time ( $\hat{t}$ ). Now, choosing  $k = \frac{1}{\beta_c}$  for simplicity, our equation becomes;

$$\frac{dH}{d\hat{t}} = H' = -HC$$

This process can be repeated for the C and R equations, yielding the dimensionless system of equations;

$$\begin{aligned} H' &= -HC \\ C' &= HC - \frac{\gamma_c}{\beta_c} C \\ R' &= \frac{\gamma_c}{\beta_c} C \end{aligned}$$

Natural units makes understanding the behaviour of a model easier due to being able to compare models which have different initial parameters. In this instance, scaling the model relative to the infection rate ensures that the time scale will encapsulate one of the most important physical properties of the system.

For readability and simplicity, from this point onward,  $\frac{\gamma_c}{\beta_c} = \gamma$ .

$$\begin{aligned} H' &= -HC \\ C' &= HC - \gamma C \\ R' &= \gamma C \end{aligned}$$

## 2.5 Fixed Points

Once a fixed point is reached, there is no possible change to the populations in the system. Fixed points can be found by equating the governing differential equations in the system to 0.

$$H' = -HC = 0 \therefore H = 0$$

$$C' = HC - \gamma C = 0$$

$$R' = \gamma C = 0, \gamma > 0 \therefore C = 0$$

Therefore, there are two fixed points;

$(0, 0, R)$ : In this scenario, every single student has dropped out of Imperial after becoming Cooked and subsequently failing.

$(H, 0, R)$ : In this scenario, although a small Cooked population was present initially, they all dropped out before their bad habits took hold of the entire campus.

The initial parameters used in the model (set as default in Wolfram Mathematica but adjustable using the sliders) are:  $H_0 = 0.95, C_0 = 0.05$ .  $C = 0.05$  is an approximation for the percentage of total students already have poor study habits upon enrollment.

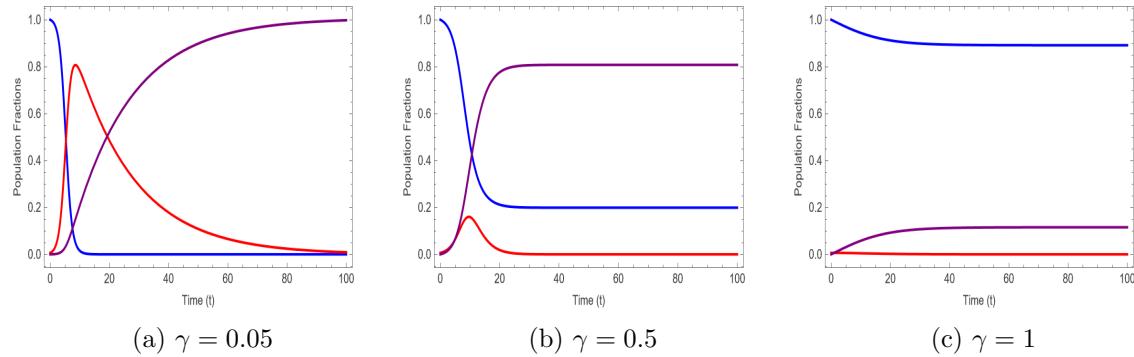


Figure 4: Population change over time at different  $\gamma$  values. (H (blue), C (red) and R (purple))

These fixed points show that co-existence is impossible, as for there to be co-existence, C must be  $> 0$ . If  $C > 0$ , then  $R' > 0$  which would result in a non-stationary system.

Both fixed points can be seen in Fig. 4, where the populations level off.

As total population in the system is conserved, and R does not influence H or C, the system can be reduced to 2 dimensions.

$$H' = -HC$$

$$C' = HC - \gamma C$$

## 2.6 Stability Analysis

Stability analysis shows what happens when state variables of the system start close to a fixed point. Practically, this will help us to understand whether if we manage to reach the (H, 0, R) state in which the healthy students have prevailed, will the re-introduction of one or two Cooked students cause mass drop-outs again? Or will the Cooked population quickly decline back to 0?

To carry out stability analysis for the identified fixed points, a Jacobian matrix will be formed to represent the system.

$$J(H, C) = \begin{bmatrix} -C & -H \\ C & H - \gamma \end{bmatrix}$$

At the fixed point (0, 0, R);

$$\det(J) = 0$$

Since the determinant of the matrix is zero, one of the eigenvalues must also be zero. The zero eigenvalue corresponds to a direction in which disruptions to the population do not get restored to the fixed point or blown up.

**Note to self:** I don't think this is quite right here

## 2.7 Eigenvalues

I think this section needs some fixing, I'm not sure what I've fucked up.

Solving the following equation for the eigenvalues;

$$\det(J - \lambda I) = \begin{vmatrix} -C - \lambda & -H \\ C & H - \gamma - \lambda \end{vmatrix} = 0$$

$$\begin{aligned} (-C - \lambda)(H - \gamma - \lambda) - (-H)(C) &= 0 \\ (C + \lambda)(H - \gamma - \lambda) - HC &= 0 \\ (-1)\lambda^2 + (H - \gamma - C)\lambda - C\gamma &= 0 \end{aligned}$$

Then, using the quadratic formula;

$$\lambda_{2,3} = \frac{(H - \gamma - C) \pm \sqrt{(H - \gamma - C)^2 - 4(C\gamma)}}{2}$$

From here, the fixed point(s) are substituted into the equation to find the remaining eigenvalues for each case.

**Note to self:** Simplify the quads a bit too many zeros!

For case (H, 0, R);

$$\begin{aligned} \lambda_{2,3} &= \frac{(H - \gamma - (0)) \pm \sqrt{H^2 - 2H(0) - 2H\gamma + (0)^2 - 2(0)\gamma + \gamma^2}}{2} \\ &= \frac{(H - \gamma) \pm \sqrt{H^2 - 2H\gamma + \gamma^2}}{2} = \frac{(H - \gamma) \pm \sqrt{(H - \gamma)^2}}{2} \end{aligned}$$

$$\lambda_2 = \frac{(H - \gamma) - (H - \gamma)}{2} = 0, \lambda_3 = \frac{(H - \gamma) + (H - \gamma)}{2} = H - \gamma$$

To summarise:

Fixed Point	$\lambda_1$	$\lambda_2$
(0, 0, R)	0	- $\gamma$
(H, 0, R)	0	H- $\gamma$

This model will always converge to a state with form (H, 0, R), for (0, 0, R), H = 0.

For  $\lambda_2$  at (H,0,R), the sign of the eigenvalue depends on the values of H and  $\gamma$ . Once at the fixed point (H, 0, R), stability is dependent on the manner in which the perturbations end up arising. If  $\gamma > H$ ,  $\lambda_2 < 0$ , and perturbations in C decay, resulting in the system returning to a disease free state quickly. Otherwise if  $\gamma < H$ , the system will become unstable as  $\lambda_2 > 0$  and perturbations in C will grow.

In the case where H = 0, (0, 0 , R),  $\lambda_2$  is always negative, as  $\gamma > 0$ . From this we can interpret that there is stability in this direction, and any small perturbations will quickly decay.

Due to the presence of zero-eigenvalues we can interpret that the behaviour in these directions is driven by non-linear terms in the system. This results in the presence of a family of fixed points on the line (H, 0, R). Whatever value of H the system has, as soon as the population of unhealthy students (C) = 0, the system stops changing. To better analyse the system behaviour, we will next carry out nullcline analysis.

## 2.8 Phase Space Analysis

Nullclines are lines where the rate of change of variables is zero. For the current HCR model, these are aligned with the x or y axis of the phase space graph. This is due to the presence of fixed points—when the healthy student population reaches the fixed point value, the nullcline is reached and only the Cooked student population will decline, with no change in the Healthy student population.

Phase space analysis gives us a way to see these nullclines, and therefore examine the way in which the population of Healthy students and Cooked students will evolve over time from different states.

The behaviour of the system is governed by the removal rate  $\gamma$ , and the initial starting conditions. Phase space diagrams show the behaviours over a range of initial population conditions, so the graphs are generated solely from differing values of  $\gamma$ .

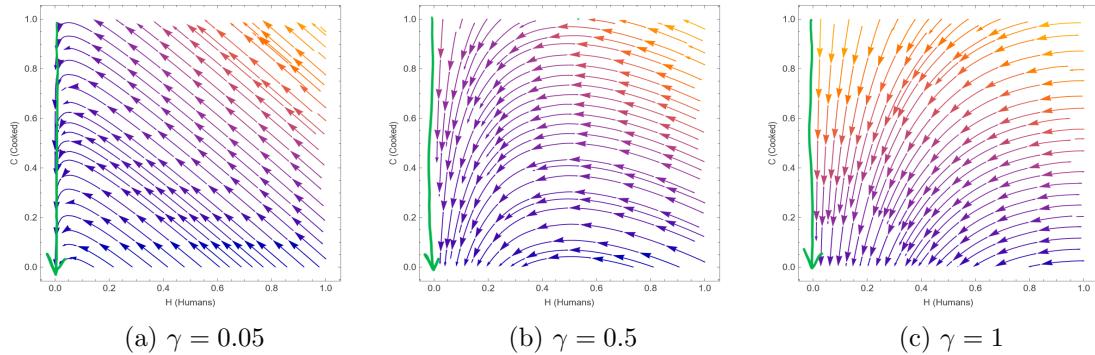


Figure 5: Phase plots ( $H$  vs.  $C$ ) — at different  $\gamma$  values.

Three graphs have been generated, one with a low removal rate (0.05), one with a mid-level removal rate (0.5) and one with a high level removal rate (2.0). To restate, these numbers represent the rate at which the Cooked naturally enter the removed population, without the influence of any external factors. Varying  $\gamma$  does not change the location of the vertical nullcline, simply the speed at which the arrows converge on it. Varying  $\gamma$  also has an effect on  $H$ , as can be seen in Fig.5 (a-c), the arrows pointing into the x-axis all converge between  $H = 0$  and  $H = \gamma$ .

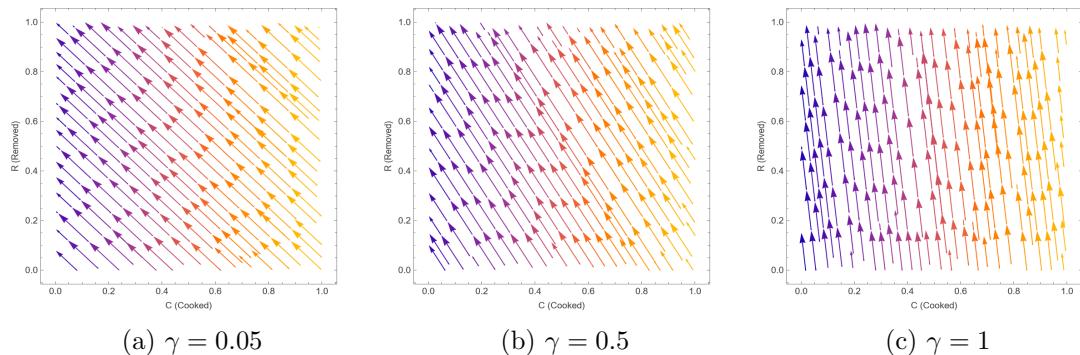


Figure 6: Phase space ( $C$  vs.  $R$ ) — at different  $\gamma$  values.

Phase plots for (C vs. R) show the influence of  $\gamma$ , with increasing  $\gamma$  values returning steeper arrows representing rapid removal of C to R. This corresponds to Cooked students dropping out of Imperial faster.

In both sets of phase space diagrams in Figure 5 & 6, the arrows point toward a decline in both H and C populations, representing the healthy students gradually being Cooked and the Cooked gradually dropping out.

$\gamma = 0.05, \mathbf{0}, \mathbf{0}, \mathbf{R}$ : In Figure 5,6 (a), we can see that with a very low removal rate, the Cooked are very successful at Cooking the remaining students on campus. By the time that almost all of the healthy students are Cooked, the Cooked have only just begun to enter the Removed population. It can be seen that all arrows are converging around the point (0,0).

$\gamma = 0.5$ , **Small H,0,R:** With a mid-level removal rate In Figure 5,6 (b), the Cooked have an initial gain in population, but the gain quickly levels out and then declines. The arrows converging onto the x-axis show that the healthy population eventually stabilises above 0 when there are no remaining Cooked present.

$\gamma = 1$ , **Large H,0,R:** In contrast, in Figure 5 (c), the Cooked population is in constant gradual decline over time. This decline in Cooked is slowed by larger initial values of Healthy population, but always ends with the Cooked population converging toward 0. Also, it can be seen that the healthy population stabilises at a wider range of values (0 to  $\gamma$ ) than in Fig 5. (c).

## 2.9 Control Parameter / Threshold

$R_0$  is the basic reproduction number, defined as the expected number of Healthy students a single Cooked student is able to influence before dropping out, and can be expressed as;

$$R_0 = \frac{\beta_c}{\gamma}$$

This also represents the threshold at which the cooked student population either grows or declines.

$R_0 < 1$ : Each cooked student, on average produces less than one student with poor habits before they drop out, leading the final healthy student population greater than the removed population.

$R_0 = 1$ : Each cooked student influences one healthy student on average.

$R_0 > 1$ : Each cooked student is able to influence more than one healthy student, leading the final healthy student population to be less than than the removed population.

## 2.10 Model Evaluation

This model gives us helpful insight into important factors influencing the outcome of the Cooked take-over, such as the drop-out rate  $\gamma$  and transmission rate  $\beta_c$ . It is simple and clearly interpretable. This model highlights that increasing  $\gamma$  is an effective strategy for stabilising the Healthy student population.

However, it also implies that the best hope for Imperial is for all of the Healthy students to leave campus and wait for the Cooked to drop out in isolation before returning to campus. This model does not model any agency for Cooked student trying to improve their study habits or intervention measures.

## 3 First Alteration

By modelling Cooked student's efforts to improve their study habits, a way for the Cooked to re-enter the Healthy student population is created.

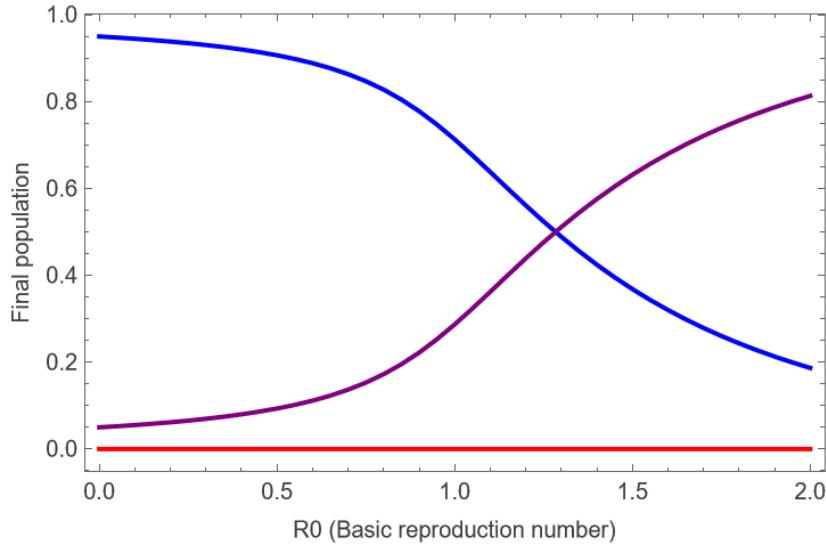


Figure 7: Effect of varying reproduction rate ( $R_0$ ) on final populations.

### 3.1 Governing Equations

A new parameter  $\alpha$ , can be defined as the 'recovery' rate, representing how quickly the Cooked students can improve their habits and re-enter the Healthy student population. The bigger  $\alpha$  is, the more Cooked students successfully form new Healthy habits.

$$\begin{aligned} H' &= -\beta_c HC + \alpha_c C \\ C' &= \beta_c HC - \gamma_c C - \alpha_c C \\ R' &= \gamma_c C \end{aligned}$$

With parameters;

$\alpha$ : Self-recovery rate of Cooked students

The system diagram can also be updated to show this.

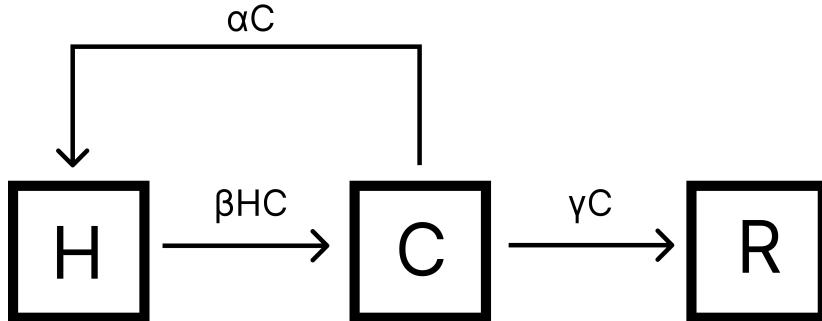


Figure 8: Updated HCR model 2.0.

### 3.2 Natural Units

The same process from earlier will be used to non-dimensionalise the equations in the model, yielding the dimensionless system of equations;

$$H' = -HC + \alpha C, \quad C' = HC - \gamma C - \alpha C, \quad R' = \gamma C$$

### 3.3 Fixed Point Analysis

Fixed point analysis will not be repeated as it shows that the location of the equilibrium points remains unchanged—(H, 0, R).

$$H' = -HC + \alpha C = 0$$

$$C' = HC - \gamma C - \alpha C = 0$$

$$R' = \gamma C = 0$$

Instead, the reduced subsystem will be examined. Similarly to in Section 1, H and C determine the entire behaviour of the system, and the system can therefore be re-written two dimensionally as such;

$$H' = (\alpha - H)C$$

$$C' = (H - \gamma - \alpha)C$$

**Slowing the Spread of Bad Habits:** The addition of the  $\alpha$  terms slows down the rate at which the healthy population gets Cooked, and also speeds up the rate of reduction of the Cooked population. This can be seen in the gradient of the population curves in Fig. 8. As  $\alpha$  is increased, the gradients become steeper, reaching the fixed point faster.

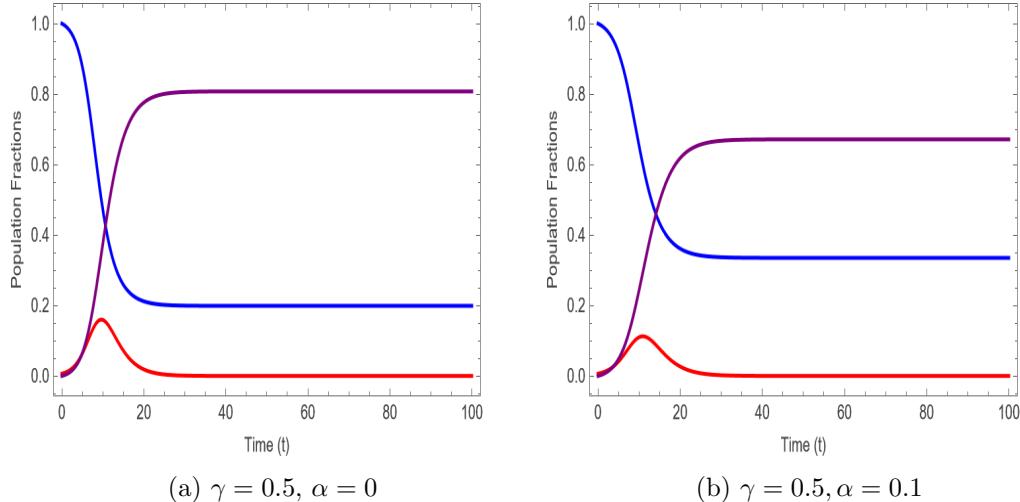


Figure 9: Population change over time with student-self improvement. (H (blue), C (red) and R (purple))

The final populations of H and R are altered by a shift  $\alpha$ , visible in the time population graphs (Fig. 9). H is shifted up, R down, and additionally the peak of C is shifted down by  $\alpha$ .

### 3.4 Nullcline Analysis

Once again, to indentify the nullclines;

$$H' = (\alpha - H)C = 0 \therefore H = \alpha$$

$$C' = (H - \gamma - \alpha)C = 0 \therefore H = \alpha + \gamma$$

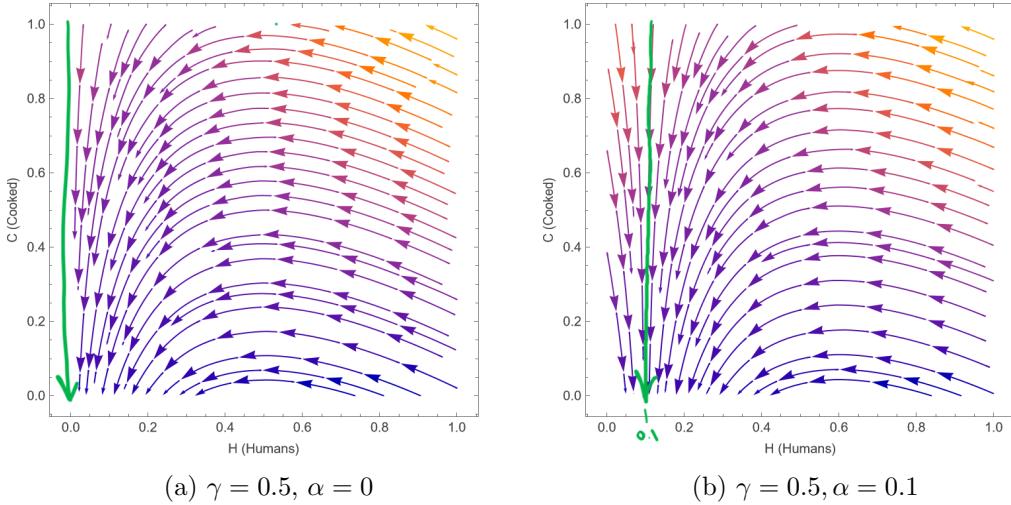


Figure 10: Phase space (H vs C) plots with nullclines shown in green.

Another way to interpret the effect on the model by student self-recovery is to examine the change in the nullclines. Previously, the nullclines were at  $H = 0$  and  $H = \gamma$ . The introduction of the alpha terms have shifted the nullclines by a factor of  $\alpha$ . The nullclines are now at  $H = \alpha$  and  $H = \gamma + \alpha$  (Fig. 10).

### 3.5 Control Parameter / Threshold

It can also be seen from  $C' = (H - \gamma - \alpha)C$ , that in a Healthy only population, if  $H > \gamma + \alpha$  then  $C' > 0$  and the Cooked population will grow, and if  $H < \gamma + \alpha$  then  $C' < 0$  and the Cooked population will decay.

$R_0$ , our basic reproduction number is therefore modified;

$$R_0 = \frac{\beta_c}{\gamma + \alpha}$$

The threshold value remains the same at  $R_0 = 1$ , however the introduction of student-self improvement (represented by  $\alpha$ ), reduces the value of  $R_0$  for fixed beta and  $\gamma$ . This lowers the expected number of Healthy students influenced by each individual Cooked student.

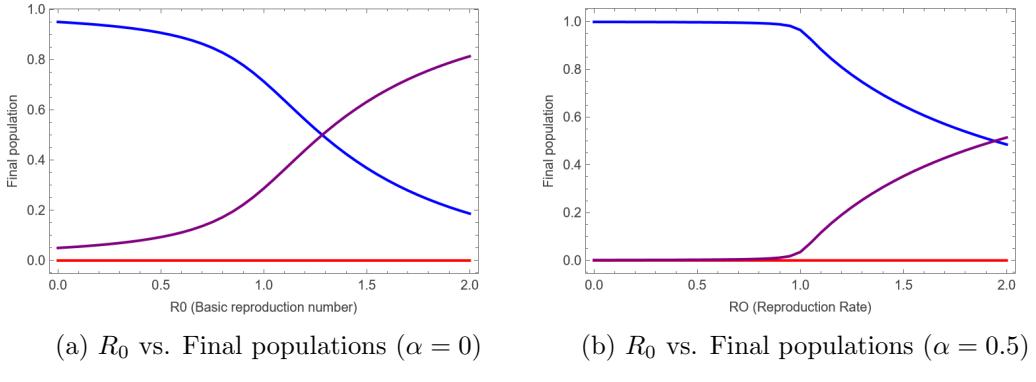


Figure 11: Effect of varying reproduction rate ( $R_0$ ) on final populations.

This shows that if the basic reproduction rate  $R_0$  is less than 1, then all bad habits will die out without requiring further intervention. In this circumstance, any small introduction of unhealthy habits decays naturally.

However, if  $R_0 > 1$ , poor academic habits spread faster than unhealthy students are forced to drop out, leading to growth in the Cooked population and heading toward mass drop-outs.

### 3.6 Model Evaluation

Allowing re-entry to the Healthy student population removed the unrealistic expectation that the decline of the student population is irreversible and inevitable.

Where previously the only option for Cooked students was to passively wait until they had to drop out—the inclusion of  $\alpha$  accounts for struggling students actively trying to improve their own study habits.

However, in the real world, struggling students are not left to struggle alone, and there are interventions and resources that can be provided by Imperial to support unhealthy students to improve their habits. As well as this, Imperial provides mitigating circumstances which prevent many students from having to retake modules, failing, or dropping out.

## 4 Second Alteration

Imperial offers study skills workshops and well-being meetings to empower students to improve how they manage their academic workload. In an attempt to provide more support to students in bas circumstances, Imperial also provides students with mitigating circumstances and the option to take a year out if things get really bad.

### 4.1 Governing Equations

As opposed to Cooked students dropping out entirely, the introduction of mitigating circumstances ( $\delta$ ) means that the group of would-be drop-outs are receiving support to prevent them from dropping out. These interventions ensure the overwhelming majority of Imperial students can remain in Imperial, however students that are given assistance to remain at

Imperial do not return as Healthy as they were upon enrolment. Instead, students that were in the Removed group re-enter the Cooked student population.

With the addition of mitigating circumstances, the system of equations becomes;

$$\begin{aligned} H' &= -\beta_c HC + \alpha_c C \\ C' &= \beta_c HC - \gamma_c C - \alpha_c C + \delta_c R \\ R' &= \gamma_c C - \delta_c R \end{aligned}$$

With parameters;

- $\alpha$ : Recovery rate of Cooked students, increased by the introduction of well-being meetings.
- $\delta$ : Representing Mitigating Circumstances, shows the rate of would-be drop-outs re-entering the Cooked population.

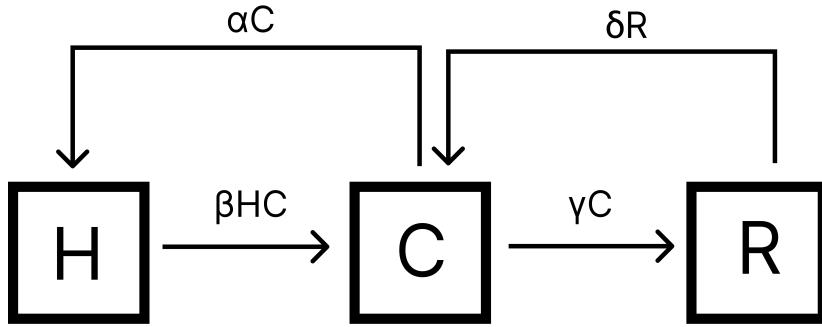


Figure 12: Introduction of mitigating circumstances and interruption of studies.

## 4.2 Natural Units

Repeating the same non-dimensionalisation process as in the previous 2 models, the new system of equations is;

$$\begin{aligned} H' &= -HC + \alpha C \\ C' &= HC - \gamma C - \alpha C + \delta R \\ R' &= \gamma C - \delta R \end{aligned}$$

## 4.3 Fixed Point Analysis

In the previous sections, transition to R was an irreversible 'sink.' Now that R flows back into C, the sink is gone, and a non-trivial (endemic equilibrium) fixed point is present.

Equating our equations for H and R to 0;

$$C(\alpha - H) = 0$$

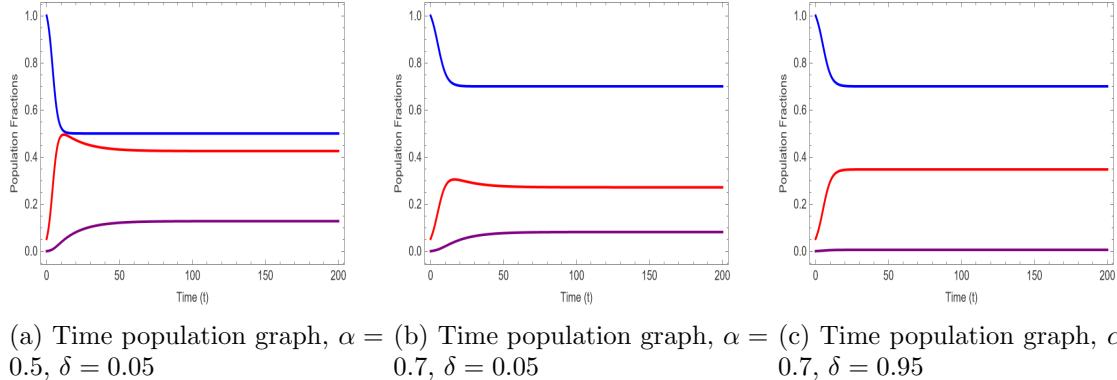
$$R = \frac{\gamma}{\delta} C$$

For the endemic equilibrium,  $C$  does not = 0, therefore,  $H = \alpha$ . Substituting these into the total population constraint ( $H + C + R = 1$ );

$$\begin{aligned}\alpha + C + \frac{\gamma}{\delta}C &= 1 \therefore C(1 + \frac{\gamma}{\delta}) = 1 - \alpha \\ \therefore C &= \frac{1 - \alpha}{1 + \frac{\gamma}{\delta}}, R = \frac{\gamma}{\delta}(\frac{1 - \alpha}{1 + \frac{\gamma}{\delta}})\end{aligned}$$

This gives us an equilibrium equation, so long as  $0 < \alpha \geq 1$ ,  $C \geq 0$ ,  $R \geq 0$ , and  $0 \leq H \leq 1$ . If  $\alpha \geq 1$ , the top of the fraction would become negative.

This demonstrates that for Cooked students to exist in equilibrium ( $C \neq 0$ ), the population of healthy students becomes fixed at  $H = \alpha$ . Increasing the 'recovery rate' directly increases the steady-state number of healthy students.



(a) Time population graph,  $\alpha = 0.5$ ,  $\delta = 0.05$       (b) Time population graph,  $\alpha = 0.7$ ,  $\delta = 0.05$       (c) Time population graph,  $\alpha = 0.7$ ,  $\delta = 0.95$

Figure 13: Impact of increasing  $\alpha$  from 0.5 to 0.7,  $\delta$  from 0.05 to 0.95.

The effects of varying  $\alpha$  and  $\gamma$  can be seen in Fig. 13 (a-c).

**Varying  $\alpha$ :** Once again a shift by  $\alpha$  is observed in the final student populations (Fig. 13 (a,b)), increasing the Healthy student population and decreasing the Cooked population. The peak of Cooked student population is also suppressed, as well as the final number of students receiving mitigation (Removed).

**Varying  $\delta$ :** Increasing  $\delta$  does not effect the final population of Healthy students, which remains determined by  $\alpha$ . However, increasing  $\delta$  does reduce the number of students in the final Removed population and raise the final Cooked population. This shift aligns with the interpretation of  $\delta$  as the rate at which struggling students can be supported by mitigating circumstances, increasing the number of Removed students that can re-enter the Cooked student population.

#### 4.4 Stability & Phase Space

Recalculating the Jacobian for the reduced 2D system and evaluating at the endemic equilibrium ( $H = \alpha$ ,  $C = \frac{1-\alpha}{1+\frac{\gamma}{\delta}}$ ) yields;

$$J(H, C) = \begin{bmatrix} -C & 0 \\ C - \delta & -(\gamma + \delta) \end{bmatrix}$$

Therefore the eigenvalues are:

$$\lambda_1 = -C = -\frac{1-\alpha}{1+\frac{\gamma}{\delta}}, \lambda_2 = -(\gamma + \delta)$$

Since all of the parameters are positive and  $\alpha < 1$ , both eigenvalues are both real and negative. This shows that the endemic equilibrium is a stable node.

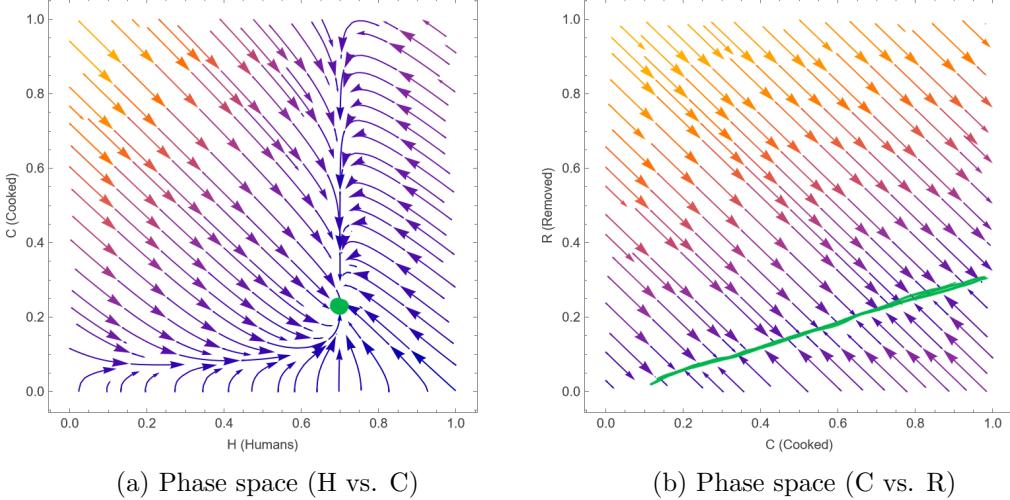


Figure 14: Phase space plots showing sinkhole (H vs. C) and equilibrium line (C v. R) in green.

This behaviour can also be seen in the phase space diagrams (Fig. 14(a)), where the arrows converge inward toward the green dot representing the endemic equilibrium point. Within the relevant region, this equilibrium point functions as a global attractor. This means that the system converges to this point regardless of the initial number of Cooked students with poor study habits. In predator-prey systems, the feedback loops often result in oscillatory behaviour, however the feedback in this model is overdamped, preventing cyclic outbreaks. Regardless of the number of initial students with poor habits, the system converges toward a the endemic equilibrium state, in which a constant population of students present that maintain their bad study habits.

Setting  $C' = 0$  permits solving for the CR-nullcline.

$$R' = \gamma C - \delta R = 0 \therefore R = \frac{\gamma}{\delta} C$$

$$C' = HC - (\gamma + \alpha)C + \delta R = 0 \therefore R = \frac{\gamma + \alpha - H}{\delta} C, H = \alpha \therefore R = \frac{\gamma}{\delta} C$$

This line (shown in green Fig 14.(b)) represents the line in which endemic equilibrium is reached, where  $H = \alpha$ .

#### 4.5 Control Parameter / Threshold

Another graph can be created to visualise the final populations when varying reproduction rate. Since  $\delta$  influences the transition of Removed students re-entering the Cooked population,

basic reproduction number remains the same as in the previous section. This is due to  $R_0$  measuring an initial growth of C when starting from a majority Healthy population, where  $R \approx 0$ .

$$R_0 = \frac{\beta_c}{\alpha + \gamma}$$

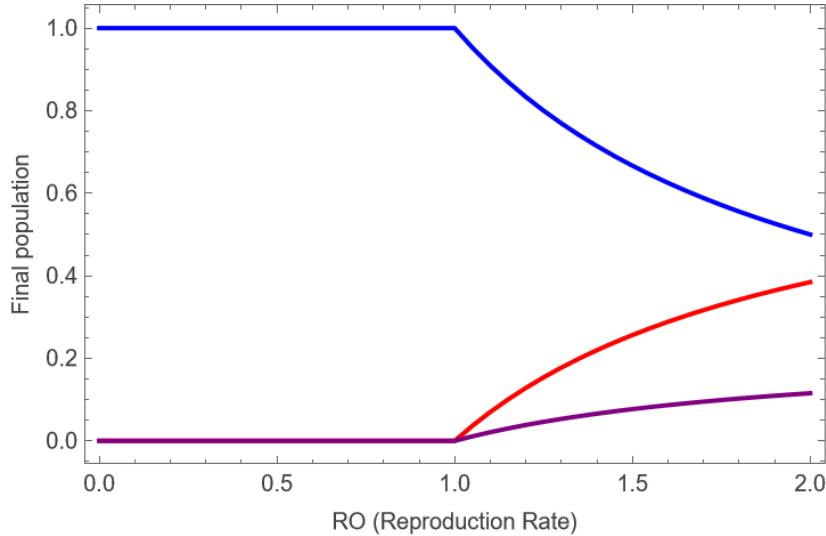


Figure 15: Effect of varying reproduction rate ( $R_0$ ) on final populations.

$\delta$  effects the final equilibrium populations and the speed that the system converges, but does not change the bifurcation point, because the basic reproduction number remains unchanged.

The system always has a fixed point with no Cooked students present ( $C = 0$ ), which is stable when  $R_0 < 1$ . As  $R_0$  increases past the critical threshold ( $R_0 = 1$ ), a transcritical bifurcation occurs (as seen by a sharp curve upward in Fig. 15.) The state with no Cooked students loses stability and a new stable state emerges. This state is endemic equilibrium, in which  $H = \alpha$ , and the proportion of students in the Cooked and Removed populations is dependent on  $\gamma$  and  $\delta$ . In this state, the Cooked students can co-exist indefinitely, alongside the Healthy students.

#### 4.6 Model Evaluation

To investigate the robustness of the model, a Monte Carlo ensemble simulation was carried out.

The Monte Carlo ensemble simulation shows the results of 50 runs with randomized parameters ( $\pm 3\%$ ). The thickness of the resulting population bands indicates the sensitivity of the model. Generally, the the bands follow follow the same general curve shown previously, demonstrating that the model is somewhat stable—small changes in parameter estimates will not lead to wildly different predictions about the impact of bad habit spread.

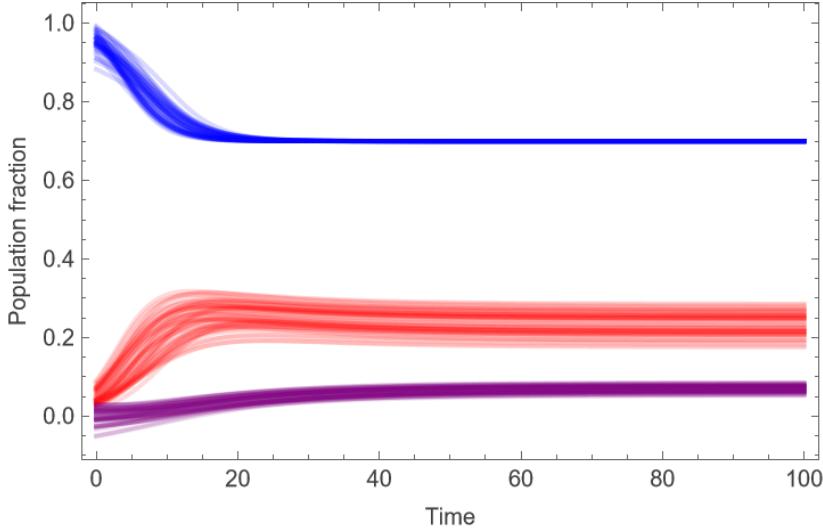


Figure 16: Running simulation x50 with random noise ( $\alpha = 0.7, \delta = 0.05.$ )

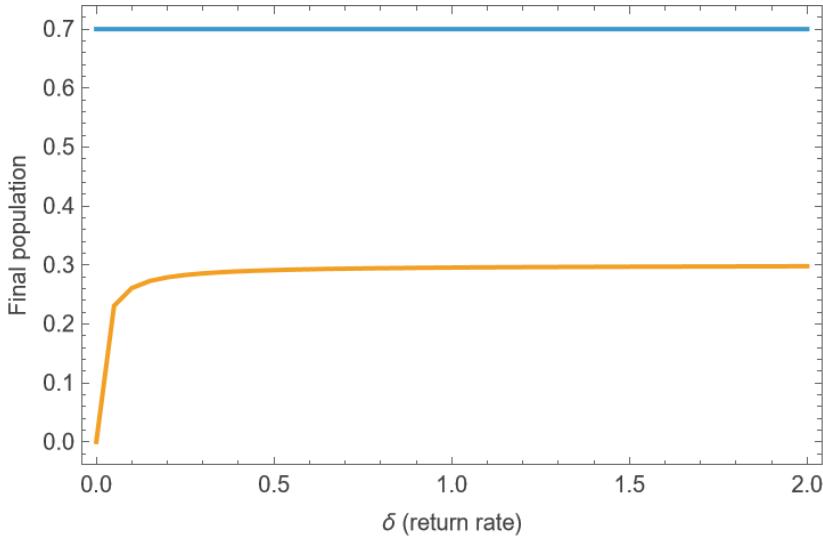


Figure 17: Sensitivity analysis of 'Mitigating Circumstances'

As the recovery rate  $\delta$  increases, the final population of unhealthy Cooked students also increases. This correlates well to real-world application, as the more Imperial is able to encourage struggling students to keep going without actually improving their long-term situations, the more unhealthy students will be stuck in a state of needing repeated support interventions.

Although the addition of mitigating circumstances and well-being meetings had the desired effect of eradicating drop-outs, the pay-off is that the cooked student population now stabilises above zero. No students are dropping out of Imperial, but a high proportion of students are left struggling.

This poses a question—is it better to have a student population that is completely healthy

but at the expense of a large number of drop-outs? Or is it better to have a population of struggling students that are given mitigations to prevent dropping out, but perpetuating their struggle?

At a high-pressure university such as Imperial, this constant population of struggling students is very representative of reality. As discussed earlier, up to 40% of students partake in all-nighters and other unhealthy studying habits[3].

## 5 Conclusion

To summarise;

**Original HCR Model:** Demonstrated that without any intervention, the unhealthy habits of the cooked students naturally burn out alongside them. These high drop-out rates sacrifice large percentages of the student population to ensure that anyone with bad habits is kicked out/drops out and does not influence other students. The only two possible outcomes were total student eradication, or only students with healthy habits prevailing, dependent on the value of  $\gamma$ .

**Self-improvement of Students:** The addition of terms to model the students' drive to improve their own study habits delayed the collapse of the student population, but could only prevent collapse if the number of students with poor habits was below a critical threshold  $(1 - (\gamma + \alpha))$ .

**Mitigating Circumstances & Well-being Meetings:** The introduction of more extreme support measures created a closed loop system. This prevented total collapse of the student population, but at the cost of the permanent presence of students with unhealthy habits.

So, my recommendations to improve overall student well-being and stop all the students getting cooked?

**Maximise Well-being Resources:** The model reveals that maximising  $\alpha$ , representing students' self-improvement directly determines the final Healthy student population. Therefore, maximising  $\alpha$  is far more valuable than maximising  $\delta$  if the goal is elimination of unhealthy habits. Section 4.3 demonstrates that the Healthy population is strictly limited by  $\alpha$ . Therefore, to ensure that at least 80% of the student body maintains healthy study habits, the university must sufficiently fund long-term interventions for students. This could include well-being meetings, mentoring, and study skills building workshops. Only increasing  $\delta$  (mitigation of dire circumstances) without increasing  $\alpha$  may result in an academic environment rife with poor study habits—high student retention but at the cost of low student well-being & academic performance.

For future improvements to the model;

**Improve Robustness:** Although Monte Carlo ensemble simulations were performed, the model is sensitive to larger fluctuations in initial student numbers. Noise levels above 0.03 cause instability, which is significant when modelling real-world scenarios. According to Imperial statistics, although the percentage change of students over one year is only 3%, the increase over the last five years is over 14%[2].

**Model Departmental Spread:** It was assumed that the student mix was homogenous, however a more realistic model would consider that students tend to frequent their departments and a few key shared spaces. This is another vital modelling process in pandemic modelling in which frequency of central location visitation is limited, similar to lockdown during COVID-19.

**Calibrate Against Real Data:** Real-world approximations for parameters such as the transmission rate  $\beta_c$  are difficult to determine, as there are many factors that effect change in human behaviour. According to studies, around 10-40% of students at university display unhealthy study habits such as all-nighters[3]. After conducting a small survey with current Imperial students, it was found that in an average group project size of 5, it would take 2 or more students with poor study habits to influence the entire group. A binomial distribution was then used to calculate the probability that in a group of 5, there are at least two students with unhealthy habits. This yields a probability of 26.27%. Benchmarking parameters against other existing infections/university statistics would increase both the accuracy and usefulness of the model.

**Event Modelling:** Events such as deadlines and holidays have a massive impact on the well-being of the overall student body and would play a vital role in a model of Imperial student wellness. Implementing this could prove useful for Imperial to work out how to minimise deadline stress and maximise impact of holiday rest.

**Custom Student Parameters:** This model assumed a homogenous mix of identical students, however in reality there are some types of students who have more influence over other's behaviour, those who are naturally more likely to become Cooked, build poor habits, or that are more susceptible to being influenced by others. Could interventions take priority for students more susceptible than others?

## References

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