Dynamics of Prion Proliferation Under Combined Treatment of Pharmacological Chaperones and Interferons

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Abstract

Diseases such as mad cow disease in bovines, chronic wasting disease in cervids, and Creutzfeldt-Jakob disease in humans are incurable illnesses caused by prions. Prion diseases are caused when the prion protein PrP^{C} misfolds into PrP^{Sc} , which is capable of inducing further misfolding in healthy PrP^{C} proteins. Recent *in vivo* experimental results have shown that pharmacological chaperone treatment can be used to prevent this conversion, where the pharmacological chaperones act as a short-term "vaccine" against the PrP^{Sc} proteins. A second strategic approach uses interferons to decrease the concentration of PrP^{Sc} . In this work, a model using a non-linear system of ordinary differential equations is constructed to model how these two treatments slow the proliferation of prions in the brain. Through this work it was found that interferons have a greater effect on the prion population over time, but that the pharmacological chaperones begin to effect the system earlier. This information can guide future prion experiments and inform potential treatment protocols.

1 Prion Diseases and Possible Treatments

Prions cause incurable diseases that cause irreversible neurodegeneration in the brain. Once infected by prion disease, a person has months, years, or even decades of feeling normal before symptoms appear. Once symptoms begin, the brain is slowly becoming spongy, deteriorating where the prions accumulate [28]. This neurodegeneration causes a host of crippling symptoms, like dementia, uncontrollable spasmodic movements (present in Creutzfeldt-Jakob disease), or the inability to sleep (as in fatal familial insomnia). These diseases are always fatal [28]. While individuals can be infected by outside sources, such as contaminated meat in the case of mad cow disease, prion diseases can occur spontaneously [6]. Several of these fatal prion diseases are scrapie in sheep; mad cow disease in bovines; chronic wasting disease in cervids and kuru, and Creutzfeldt-Jakob disease in humans [6]. These diseases affect the brain, causing neurons to die; this eventually leads to the death of the individual, as the brain cannot perform its essential functions [6].

Currently, prion diseases have no cure [36], so any strides towards a treatment are important. Even though prion diseases are far from commonplace, they are fatal and kill hundreds of people every year. In 2017 alone, over 500 people in the United States died from Creutzfeldt-Jakob disease [7]. Further, the study of prion diseases has implications for other neurodegenerative diseases, such as Alzheimer's and Parkinson's, as these illnesses are very similar to prion diseases. That is, they involve the loss of function of the PrP protein which negatively affects the brain's function [13].

Little is known about the specific functions of the prion protein, PrP; however, it is known that PrP slows neuronal apoptosis (cell death) [34]. Prions are created when the protease resistant protein (PrP) misfolds [28]. These proteins appear normally in mammalian brains [28]. The mechanisms of this folding error are not yet fully understood. However, it is known that misfolded proteins can cause other properly folded PrP to form into a prion [28]. This correctly folded form is sometimes called PrP^{C} (C for "cellular" [28]). Problems begin when this protein folds into an isoform¹, called PrP^{Sc} (Sc stands for "scrapie" [28]). Prions have no DNA or RNA themselves, so they go against the central dogma of biology because they are still able to replicate by inducing further misfolding [28]. The "protein only hypothesis" proposes that prion replication happens without the involvement of nucleic acid [28]. When the "good protein," PrP^{C} , folds normally, its folded form is rich in α -helices. If PrP folds into a β -sheet-rich form instead of one rich in α -helix structures, it forms PrP^{Sc} and thus becomes a prion [38].

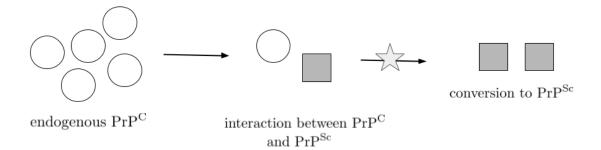


Figure 1: The Heterodimer model shows how one monomer of misfolded protein, PrP^{Sc} , converts a monomer of PrP^{C} [39].

There are two common hypotheses used to describe prion spread. The first hypothesis is the heterodimer² model. This model assumes that when a PrP^{Sc} protein comes into contact with a PrP^{C} protein, the prion unfolds the healthy protein and acts as a template to turn the PrP^{C} into PrP^{Sc} (see Figure 1) [18]. This simple model, however, does not include the experimental fact that prions form polymers: chains of PrP^{Sc} monomers. The second hypothesis is called the nucleated polymerization model, and it studies chains of prions and how the chain length varies [27]. When a chain of prions infects a new monomer, it adds the PrP^{C} to the chain, causing it to misfold, and the chain grows by one monomer. However, if the chain breaks, there are two options. First, the polymer can break into two smaller polymers; second, if one of the polymer's length is below a certain threshold, it dissociates into monomers, [23]. In this model, the monomers are not infectious by themselves, but

¹One gene can form proteins that differ in both structure and composition; these different expressions are called isoforms [1]

²A heterodimer is protein composed of non-identical monomers [39].

they can join an existing polymer. This work will take into consideration the polymerization model (see Figure 2).

These replication models are usually in one or two spatial dimensions. The two-dimensional models study prion aggregations [20], and they have explained how incubation times and inoculation doses are highly correlated. Essentially, the period before the symptoms appear is related to the amount of prions going into the brain [20]. The one-dimensional models treat PrP^{Sc} as fibrillic structures that can add new monomers on either end of a chain. This has not only been experimentally studied [4] but also it has been widely analyzed mathematically [23, 17, 26, 8].

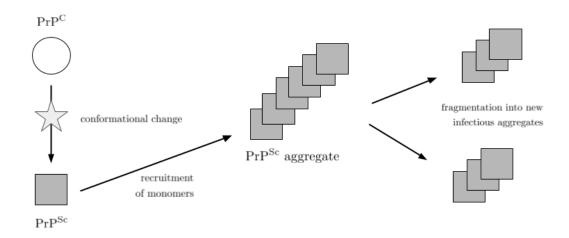


Figure 2: The polymerization model describes a prion proliferation hypothesis in which PrP^{S_c} monomers form polymer chains. These aggregates are infectious and actively recruit free endogenous PrP^{C} protein monomers that have undergone conformational change into PrP^{S_c} monomers. Large PrP^{S_c} polymer chains eventually become unstable and break into new infectious units and repeat the process [18].

1.1 Possible Treatments for Prion Diseases

Recent experimental research has shown that there are possible treatments for prion diseases [2, 24, 25, 38]. These treatments can be categorized in four mechanistic ways, as described by Kamatari [14]. In Figure 3, the first mechanism (I) is a stabilization of the PrP^C structure by direct association of a molecule to prevent formation to the PrP^{Sc} isoform. Mechanism II is

an indirect association between the interfering molecule and PrP^{C} ; this blocks the interaction between PrP^{C} and PrP^{Sc} , slowing the rate at which PrP^{Sc} is able to spread. Mechanism III removes PrP^{C} from the system, and mechanism IV prevents PrP^{Sc} from proliferating by associating PrP^{Sc} with molecules other than PrP^{C} . See Figure 3 for a visual depiction of these mechanisms.

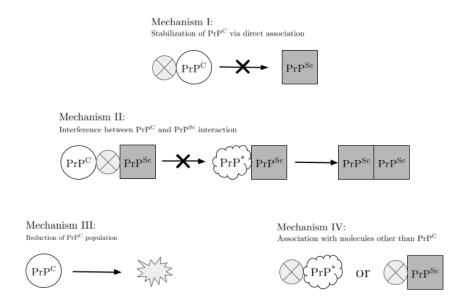


Figure 3: Four distinct methods of anti-prion treatment. Mechanism I describes the association of a molecule to PrP^{C} , causing the protein structure to stabilize and effectively preventing misfolding to PrP^{Sc} . Mechanism II also works by stabilizing the PrP^{C} structure, but the introduced molecule does not directly associate with the PrP^{C} protein. Mechanism III describes the removal of PrP^{C} from the population, and Mechanism IV describes an interaction between an interfering molecule and any other molecule that is not PrP^{C} (e.g. PrP^{Sc} or denatured PrP protein, PrP^*).

This paper will focus on Mechanism I, specifically through the use of pharmacological chaperones³. Many different types of molecules are able to act as pharmacological chaperones. Antibodies are a special class of these pharmacological chaperones. In vivo experiments have shown that antibodies can be used to block the proliferation of prions [2] by forcing the secondary structure of PrP protein into an α -helix form rather than β -sheets which are associated with PrP^{Sc} (see Figure 4). Specially-engineered molecules can be designed to bind at locations critical to the correct folding of PrP^C [22]. However, pharmacological

³Pharmacological chaperones are small, cell-permeable molecules that assist correct protein folding [21].

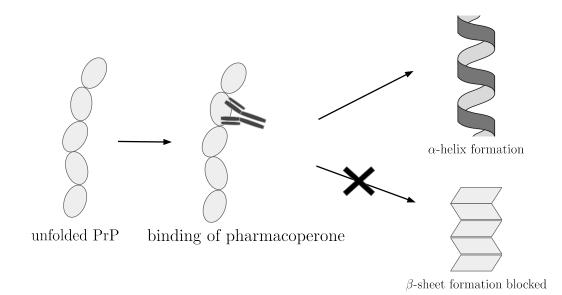


Figure 4: Pharmacological chaperones bind to unfolded PrP, forcing it to fold into α -helix-rich form and thus PrP^C instead of PrP^{Sc} which is rich in β -sheets.

chaperones tend to have a short half-life (though the specific half-life depends on the drug that is being used), which means that eventually treated PrP^{C} will become susceptible to misfolding again [32]. This suggests that we must keep the concentration of pharmacological chaperones high in order to keep the disease at bay.

Another treatment involves interferons, a part of the immune system that raises the body's immune response by signaling other proteins [5]. It has been documented that there is a naturally occurring increase in type I interferon (I-IFN) expression in a brain affected with a prion disease. This has been shown to be the case in scrapie-infected hamster and mouse brains by identifying upregulated gene expression [29, 35]. However, it is not wellknown how this innate reaction recognizes and attacks the infected proteins given that PrP^{C} and PrP^{Sc} do not differ in amino acid sequences [28]. Ishibashi et al. (2012) [11] proposed that PrP^{Sc} infection induces a response from Toll-like receptor (TLR) proteins, a class of pattern-recognizing receptors essential to the immune response signalling pathway [3]. These proteins are upstream of I-IFN, which results in a cascade of signals inducing immune response [11]. Moreover, *ex vivo* experiments in scrapie infected mice have shown that inducing I-IFN via TLR signalling reduces PrP^{Sc} concentration in the model host during the early stages of infection [12]. With these two treatments, prion formation can be slowed with pharmacological chaperones and the prion population can be significantly diminished with interferons.

In the model, a possible treatment for prion diseases is considered: a two-fold therapy, a combination of pharmacological chaperones and interferons. The findings of this research evaluated the efficacy and potency of these treatments within their safe ranges. As both pharmacological chaperones and interferons do reduce the concentration of prions, we examined whether or not a combination of sub-maximal dosages would work. This study has implications for prion disease therapies, diseases which are currently not only incurable but untreatable. This paper constructs a non-linear system of differential equations based off the previous work by Nowak et al. [23] with the addition of interferon and pharmacological chaperone treatments. We find several equilibrium points and examine their stability, as well as analyze several important aspects such as the R_0 (basic reproduction number) and the growth rate of prion proliferation. Sensitivity analysis provides insight on how much each treatment affects the population of prions. Numerical simulations provide even more insight into each treatment on its own, as well as their combination. We conclude with an examination of efficacy and potency and answer the question of how these treatments can be used to treat prion diseases.

2 Prion Proliferation Treatment Model

This model implements the polymerisation hypothesis for proliferation of prions. Two types of treatments are incorporated in this biological model. The first one consists of a dose of pharmacological chaperones which prevent prion formation based on *in vivo* experiments by Gunther et al. [10]; the second treatment consists of a dose of interferons that decreases the amount of prions in the brain [12]. Our model is based on two previous mathematical models. Masel et al. (1999) [18] used the hypothesis of nucleated polymerisation as the mechanism of proliferation. Masel et al. then established a deterministic infinite dimensional dynamical system to model the dynamics between the population of the susceptible monomers and the polymers of prions with a distinct equation to describe the population of polymers of each possible length; however, this paper did not consider any treatments. In a subsequent paper, Masel et al. (2000) [17] used a theoretical kinetic model to calculate the growth rate of protein aggregates as a function of certain drugs which blocks the ends of amyloids. However, the treatment examined in that model differs mechanistically from those examined here.

PrP^{Sc} Elongation

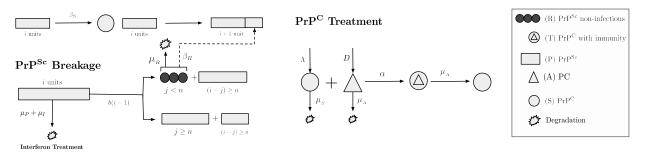


Figure 5: Detailed kinetic model that includes the nucleated polymerisation hypothesis and the incorporation of the treatments.

Figure 5 describes the kinetic model in detail. To consider the dynamics of this system, the model posed here examines what can happen with a PrP^{C} monomer. PrP^{C} monomers are naturally produced at a rate Λ . There are several paths that this monomer can take once it has joined the system. A polymer of PrP^{Sc} can convert this monomer, which happens at rate β_{S} . From there, the polymer will either split or simply grow longer. The breakage rate is $b_{i,j}$, where *i* is the length of the polymer and *j* (and thus i-j) are the lengths of the new polymers after it splits. When the polymer breaks, two events can happen. It can break in two smaller polymers, each with a length greater than *n*, and they will continue converting monomers. Otherwise, it can split into a polymer and small chain (whose length is less than the threshold *n*) which will dissociate into separate PrP^{Sc} monomers (see Figure 6 for a visual depiction of the breakage process). This monomer can join an existing polymer, occurring at rate β_{R} , but it cannot be treated with pharmacological chaperones to prevent it from rejoining a chain as the pharmacological chaperones can only act on PrP^{C} . The introduction of interferons, however, may mean that our polymer is eliminated much sooner. The interferons induce an additional death rate for PrP^{Sc} , called μ_I . If the PrP^{C} monomer is not treated with pharmacological chaperones, it can be infected. The addition of pharmacological chaperones into the system can be described by the dosage rate, D. Once a PrP^{C} monomer has been treated, it will be immune from PrP^{Sc} until either the pharmacological chaperone or the PrP^{C} monomer degrades. Degradation of pharmacological chaperones and PrP^{C} happen at rates μ_A and μ_S , respectively. These interactions are summarized mathematically in the equations section.

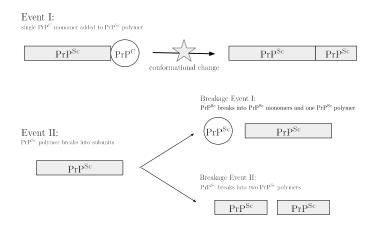


Figure 6: The possible events for a PrP^{Sc} polymer. In event I, a PrP^{Sc} polymer contacts a PrP^{C} , inducing a conformational change in which the PrP^{C} is added to the polymer. In event II, the PrP^{Sc} breaks into fragments. There are two unique breakage events. In breakage event I the PrP^{Sc} polymer breaks to create one PrP^{Sc} monomer and one PrP^{Sc} polymer. In breakage event II, the PrP^{Sc} polymer breaks to create two PrP^{Sc} polymers.

2.1 Assumptions

In addition, we implement assumptions made in the model of Masel et al. [18], such as the fact that the infection rate to convert PrP^{C} to infectious PrP^{Sc} (β_{S}) and the rate to convert non-infectious PrP^{Sc} to infectious PrP^{Sc} (β_{R}) are the same for PrP^{Sc} of all lengths $i \geq n$,

i.e. $\beta_S = \beta_R$. Also, the rate at which prions break is the same for all lengths $i \ge n$, so we let $b_{i,j} = b$. Furthermore, the death rates for $\Pr P^C$ and non-infectious $\Pr P^{S_c}$ are the same, or $\mu_S = \mu_R$ [18]. It is important to note that death rate of $\Pr P^{S_c}$, μ_P , is different. The model allows μ_I to either affect the non-infectious $\Pr P^{S_c}$ monomers or not; this is achieved by the term σ , where $0 \le \sigma \le 1$ is a probability representing the degree to which the $\Pr P^{S_c}$ monomers are affected by the interferons.

The transmission of prions follows the polymerisation hypothesis. The prion polymers are considered linear, that is, PrP^{C} can only attach to the ends of PrP^{Sc} polymers [18]. For simplicity the population of PrP^{C} is considered to be a well-mixed homogeneous system. Since the reaction that involves the misfolding of a protein are extremely fast and are measured in microseconds [16], we consider the time to fold or unfold a protein is negligible. We also make several assumptions about the treatments introduced in this paper. For the first treatment, each of the pharmacological chaperones introduced in the body binds to one PrP^{C} in order to prevent the misfolding process. It is also assumed that the pharmacological chaperone in question has high specificity, meaning we can neglect the rate at which this pharmacological chaperone binds to things which are not PrP^{C} . Moreover, the dosage of the pharmacological chaperones that is included in the brain and the dosage of interferon are assumed as constant over time.

2.2 Mathematical Model

In this model, a non-linear system of ordinary differential equations is studied. This model introduces pharmacological chaperones and interferons into a prion-infected brain. The system of infinite differential equations is presented first, followed by the reduced, closed system of six equations that will be used to study the dynamics of the prion population in an individual's brain.

$$\frac{dS}{dt} = \int_{\text{BIRTH}} -\mu_{S}S - \frac{\alpha AS}{\text{DEATH}} + \mu_{A}T - \beta_{S}S \sum_{i=n}^{\infty} P_{i}, \\
\frac{dT}{\text{INFECTION}} = -\mu_{S}T + \alpha AS - \frac{\alpha AS}{\text{INFARMACOPERONES}} + \mu_{A}T - \beta_{S}S \sum_{i=n}^{\infty} P_{i}, \\
\frac{dT}{\text{INFECTION}} = -\mu_{S}T + \alpha AS - \mu_{A}T - \mu_{A}$$

System 2.1 can be closed, and the resulting six equations are better to consider than the previous infinite system. T and A, which measure treated PrP^{C} and pharmacological chaperones respectively, are as in system 2.1. Instead of an equation for each polymer of length i, we have a class P, which counts polymers of PrP^{Sc} . P is defined as $\sum_{i=n}^{\infty} P_i$. An additional class must also be introduced in order to close this system. We define Z to be the total number of PrP^{Sc} monomers in the polymer chains, formally $Z = \sum_{i=n}^{\infty} iP_i$.

The closed system of equations is then given by:

$$\frac{dS}{dt} = \Lambda - \mu_S S - \alpha A S + \mu_A T - \beta_S S P, \qquad (2.2a)$$

$$\frac{dT}{dt} = -\mu_S T + \alpha A S - \mu_A T, \qquad (2.2b)$$

$$\frac{dP}{dt} = -(\mu_P + \mu_I)P + bZ - (2n-1)bP,$$
(2.2c)

$$\frac{dZ}{dt} = \beta_S SP + \beta_R RP - (\mu_P + \mu_I(t))Z - n(n-1)bP, \qquad (2.2d)$$

$$\frac{dR}{dt} = -\mu_R R - \mu_I \sigma R - \beta_R RP + n(n-1)bP, \text{ and}$$
(2.2e)

$$\frac{dA}{dt} = -\mu_A A - \alpha A S + D. \tag{2.2f}$$

Here, P represents the number of PrP^{Sc} polymers and Z represents the total number of PrP^{Sc} monomers within those chains. For details on how the system was closed, see Appendix B.2. The parameters used to describe the dynamics of the system are summarized in Table 1.

3 Analysis

For analysis, the model will be examined in several different cases: both treatments are being administered ($D \neq 0$ and $\mu_I \neq 0$), only pharmacological chaperones administered ($D \neq 0$ and $\mu_I = 0$), and only interferons administered (D = 0 and $\mu_I \neq 0$). The no treatments case (D = 0 and $\mu_I = 0$) is very similar to the model described in Masel et al. [18]. Any differences in analysis between this case and that which is presented by Masel et al. will be noted.

3.1 Existence of Prion-Free Equilibrium

In order to calculate the prion-free equilibrium (PFE), assume the population of prions P is zero. Then from Equations and 2.2d and 2.2e of System 2.2, the following equations are

Name	Description	Value	Units	Reference
State variables		Initial Condition		
S	$\begin{array}{c} \text{Susceptible} & \text{population} & \text{of} \\ \text{Pr}\text{P}^{\text{C}} \end{array}$	[300,750]	nM	[9]
T	PrP ^C proteins treated with pharmacological chaperones	0	nM	
P	PrP ^{Sc} chains		nM	
Z	Total of monomers for each PrP ^{Sc} polymer in the chains		nM	
R	PrP^{Sc} smaller than n	0	nM	
A	Pharmacological chaperone population	0	nM	
Parameters	<u> </u>			
Λ	Natural production of PrP ^C	[1800, 3000]	$\frac{nM}{day}$	[31]
μ_S	Degradation rate of PrP^{C}	[3,5]	$\frac{\text{day}}{1}$	[31]
μ_P	Degradation rate of PrP ^{Sc} polymers	0.047	$\frac{1}{\text{day}}$	[31]
μ_R	Degradation rate of PrP ^{Sc} monomers	[3,5]	$\frac{1}{\text{day}}$	[31]
μ_A	Degradation rate of pharma- coperones	62.0352	$\frac{1}{\text{day}}$	[19]
μ_I	Degradation rate of PrP ^{Sc} due to interferons	[0, 0.0882]	$\frac{1}{\text{day}}$	
σ	Interferons diminish R	[0,1]	_	
β_S	Infection rate of PrP^{C}	0.00292	$\frac{1}{(day)nM}$	[31]
eta_R	Infection rate of PrP^{Sc} monomers	0.00292	$\frac{1}{(\text{day})\text{nM}}$	[31]
α	Rate that a pharmacoperone binds to a PrP^{C}	0.051408	$\frac{1}{(\mathrm{day})\mathrm{nM}}$	[19]
$b_{i,j}$	Rate of breakages of the PrP^{Sc} of length <i>i</i> in a PrP^{Sc} of length <i>j</i> and $i - j$	0.0314	$\frac{1}{\text{day}}$	[31]
$n \\ D$	Minimum polymer length New dosages of pharmacologi- cal chaperones being added	$\{2,3,4,5,6\}\ [0,71500]$	$\frac{\mathrm{nM}}{\mathrm{day}}$	$[31, \ 33] \\ [15]$

Table 1: Table of parameters for the ODE's of (2.1), where $nM = \frac{nMol}{L}$

obtained:

and
$$\frac{dZ}{dt} = -(\mu_P + \mu_I)Z,$$

$$\frac{dR}{dt} = -(\mu_R + \mu_I\sigma)R.$$
(3.1)

Therefore,

$$\lim_{t \to +\infty} Z(t) = 0 \quad \text{and} \quad \lim_{t \to +\infty} R(t) = 0.$$
(3.2)

Using equation 2.2a, S satisfies the equation

$$\alpha\mu_S(\mu_A + \mu_S)S^2 + [(\mu_A + \mu_S)(\mu_A\mu_S - \Lambda\alpha) + \alpha D\mu_S]S - \Lambda\mu_A(\mu_A + \mu_S) = 0.$$

The roots of this quadratic equation are

$$S_1 = \frac{(\Lambda \alpha - \mu_A \mu_S)(\mu_A + \mu_S) - \alpha \mu_S D - \sqrt{\Delta}}{2\alpha \mu_S (\mu_A + \mu_S)},$$

and
$$S_2 = \frac{(\Lambda \alpha - \mu_A \mu_S)(\mu_A + \mu_S) - \alpha \mu_S D + \sqrt{\Delta}}{2\alpha \mu_S (\mu_A + \mu_S)},$$

where $\Delta = [(\mu_A \mu_S - \Lambda \alpha)(\mu_A + \mu_S) + \alpha \mu_S D]^2 + 4\alpha \Lambda \mu_S \mu_A (\mu_A + \mu_S)^2$. Notice that $\Delta > 0$, $\alpha \mu_S (\mu_A + \mu_S) > 0$, $-\Lambda \mu_A (\mu_A + \mu_S) < 0$ and $S_2 > S_1$ (See more details in Appendix #). Then this quadratic equation must have a positive root $S^* = S_2$. From equations 2.2b and 2.2f, we obtain the relations

$$T^* = \frac{\alpha S^* D}{(\mu_A + \mu_S)(\mu_A + \alpha S^*)} \quad \text{and} \quad A^* = \frac{D}{\mu_A + \alpha S^*}$$

That is, T^* and A^* exist and are positive, and they must have biological relevance. Therefore, the prion-free equilibrium must exist.

In our model, the PFE of the system is given by $E^* = (S^*, T^*, 0, 0, 0, A^*)$, where

$$S^* = \frac{1}{2} \left(\frac{\Lambda}{\mu_S} - \frac{D}{\mu_A + \mu_S} - \frac{\mu_A}{\alpha} \right) + \sqrt{\frac{\Lambda\mu_A}{\alpha\mu_S} + \frac{1}{4} \left(\frac{D}{\mu_A + \mu_S} + \frac{\mu_A}{\alpha} - \frac{\Lambda}{\mu_S} \right)^2}, \tag{3.3}$$

$$T^* = \frac{1}{2} \left(\frac{\Lambda}{\mu_S} + \frac{D}{\mu_A + \mu_S} + \frac{\mu_A}{\alpha} \right) - \sqrt{\frac{\Lambda\mu_A}{\alpha\mu_S} + \frac{1}{4} \left(\frac{D}{\mu_A + \mu_S} + \frac{\mu_A}{\alpha} - \frac{\Lambda}{\mu_S} \right)^2}, \text{ and} \qquad (3.4)$$

$$A^* = \frac{1}{2} \left(\frac{D}{\mu_A} + \frac{\mu_A + \mu_S}{\alpha} - \Lambda \left(\frac{1}{\mu_S} + \frac{1}{\mu_A} \right) \right) + \sqrt{\frac{\Lambda\mu_A}{\alpha\mu_S} + \frac{2\Lambda}{\alpha} + \frac{\Lambda\mu_S}{\alpha\mu_A} \frac{1}{4} \left(\frac{D}{\mu_A + \mu_S} + \frac{\mu_A}{\alpha} - \frac{\Lambda}{\mu_S} \right)^2}.$$
(3.5)

Where $P^* = Z^* = R^* = 0$ (see Appendix B.3 for details). These equilibrium values will always be real and positive, regardless of the parameter values (see Appendix B.3 for details). Additionally, notice that $S^* + T^* = \frac{\Lambda}{\mu_S}$, so when D = 0 (i.e. when no pharmacological chaperone treatment is being used) the PFE becomes $(\frac{\Lambda}{\mu_S}, 0, 0, 0, 0, 0, 0)$, the same equilibrium as in previous models which did not include treatment [18]. It can be seen then that the introduction of the pharmacological chaperone treatment lowers S^* . When $\mu_I = 0$, the prion-free equilibrium does not change with respect to the PFE.

3.2 Stability Condition of the Prion-Free Equilibrium

Theorem 3.1. The model given by System 2.2 always has a prion free equilibrium $E^* = (S^*, T^*, 0, 0, 0, A^*)$ when $R_0 < 1$, where

$$R_0 = \frac{b\sqrt{\frac{1}{4} + \frac{S^*\beta_S}{b}} - \frac{b}{2}}{\mu_I + \mu_P + b(n-1)}$$

Moreover, equilibrium E^* is locally asymptotically stable when $R_0 < 1$, and when $R_0 > 1$, E^* is unstable.

A summary of the proof of the theorem can be seen by examining the eigenvalues of the

Jacobian matrix evaluated at E^* , which are the following:

$$\begin{split} \lambda_1 &= -\mu_S < 0, \\ \lambda_2 &= -(\mu_R + \mu_I \sigma) < 0, \\ \lambda_3 &= -b(n-1) - (\mu_I + \mu_P) + b\sqrt{\frac{1}{4} + \frac{S^* \beta_S}{b}} - \frac{b}{2} \\ \lambda_4 &= -\frac{1}{2}b(2n-1) - (\mu_I + \mu_P) - \frac{1}{2}\sqrt{4S^* b\beta_S + b^2} < 0, \\ \lambda_5 &= -\frac{\alpha}{2}(A^* + S^*) - \mu_A - \frac{1}{2}\mu_S + \frac{1}{2}\sqrt{M}, \\ \text{and} \qquad \lambda_6 &= -\frac{\alpha}{2}(A^* + S^*) - \mu_A - \frac{1}{2}\mu_S - \frac{1}{2}\sqrt{M}, \end{split}$$

where $M = \alpha^2 (A^* + S^*)^2 + 2\alpha\mu_S(A^* - S^*) + \mu_S^2$. It is easy to show that $\lambda_5 < 0$ and $\lambda_6 < 0$ when M is positive. On the other hand, $\lambda_3 < 0$ if $\phi_1 < 0$ ($R_0 < 1$), where $\phi_1 = \sqrt{\Delta} + (\Lambda \alpha - \mu_A \mu_S)(\mu_A + \mu_S) - \alpha\mu_S D$, $R_0 = \frac{b(G_1 - n)}{(\mu_I + \mu_P) + b(n - 1)}$. For the details of the full proof, see Appendix B.5. As a result, equilibrium E^* is locally stable when $\phi_1 < 0$ and $R_0 < 1$, as when those criteria are met all the eigenvalues are negative.

3.3 Basic Reproductive Number

In Masel et al. [18], the basic reproductive number (R_0) of the system was found heuristically, i.e. by multiplying the rate of creation of new prions by the average lifespan of a prion (time spent in P). Table 2 shows basic reproductive values found heuristically and through a Next Generation Matrix for both Masel et al.'s system and system 2.2. Also, values found for system 2.2 are written as functions of the treatment parameters D and $/mu_I$, and $R_0(0,0) = R_0$ for both the heuristic and Next Generation Matrix values. For either system, when $R_0 = 1$ or $R_0(D, \mu_I) = 1$, the heuristic and Next Generation R_0 can be reduced to the exact same condition. This implies that these values have the same region of existence [30].

To show the heuristic expressions were determined, take $R_0^H(D, \mu_I)$. In this model, the terms which represent the degradation of prions are $(\mu_P + \mu_I + (n-1)b)P$ and the terms

Case	Heuristic	Next Generation Matrix	
Masel 1999 [18]	$R_0^{H} = \frac{\sqrt{\beta b S^* + \frac{b^2}{4}} - \frac{b}{2}}{\mu_P + b(n-1)}$	$R_0^{NG} = \frac{b(\beta S^* - n(n-1)b)}{\mu_P(\mu_P + (2n-1)b)}$	
System 2.2	$\mathbf{R}_{0}^{\mathrm{H}}(D,\mu_{I}) = \frac{\sqrt{\beta b S^{*} + \frac{b^{2}}{4}} - \frac{b}{2}}{(\mu_{P} + \mu_{I}) + b(n-1)}$	$\mathbf{R}_{0}^{\mathrm{NG}}(D,\mu_{I}) = \frac{b(\beta S^{*} - n(n-1)b)}{(\mu_{P} + \mu_{I})((\mu_{P} + \mu_{I}) + (2n-1)b)}$	

Table 2: The R₀ values for System 2.2 and previous literature found using different methods.

which represent the creation of new prions are b(Z - nP). By normalizing these terms by $\frac{1}{P}$, those terms become $(\mu_P + \mu_I + (n-1)b)$ and b(G - n), where $G = \frac{Z}{P}$ is the average length of prion polymers. The derivative of the average length of the prion polymers (G) is

$$\dot{G} = \frac{\dot{Z}}{P} - \frac{\dot{P}Z}{P^2} = \beta_S S + \beta_R R - n(n-1)b - bG^2 + (2n-1)bG.$$

Notice that near the prion-free equilibrium, $\dot{G} \approx \beta_S S^* - n(n-1)b - bG^2 + (2n-1)bG$. Therefore, this equation approximately describes behavior of a very small initial infection. The roots of this differential equation are

$$G_{+,-} = n - \frac{1}{2} \pm \sqrt{\frac{1}{4} + \frac{S^* \beta_S}{b}}$$

Because *n* is always greater than or equal to one, $G_+ = n - \frac{1}{2} + \sqrt{\frac{1}{4} + \frac{S^*\beta_S}{b}}$ is always positive. The other root, G_- , will be negative when $0 < \beta_S S^* - n(n-1)b$. This is true when $R_0^{\text{NG}}(D, \mu_I) > 0$, so G_- is always negative for biologically-relevant parameter ranges. Thus, as can be seen in the phase diagram below (see Figure 7), once an infection has been introduced to the system, the average polymer length *G* will reach a fixed value, G_+ . That means that the number of secondary infections can be represented by $R_0^{\text{H}}(D, \mu_I) = \frac{b(G_+ - n)}{(\mu_I + \mu_P) + b(n-1)}$. The difference between *n* and G_+ in the numerator represents the minimum size *n* required to be infectious. Specifically, the smallest possible value for *Z* is *nP*, where each prion chain is of its minimum length. Normalizing each of these terms shows that the minimum size of the average length *G* is *n*, so $R_0^{\text{H}}(D, \mu_I) > 0$ for biologically significant parameter values.

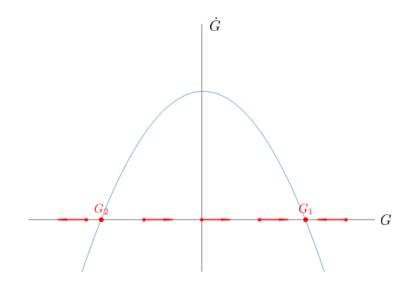


Figure 7: Phase plane for G, the average polymer length. G_+ and G_- are the fixed points of G. G_+ is a stable value for the average polymer length.

Next take the value found using a Next Generation Matrix. The secondary infection rate representing population changes in the number of monomers in prion chains (Z) is

$$\underbrace{(\beta S^* - n(n-1)b)}_{\text{Rate of new PrP}^C \text{ infected}}_{\text{by a single prior chain}} \underbrace{\frac{1}{((\mu_P + \mu_I) + (n-1)b)}}_{\text{Average time spent in P}},$$

and the secondary infection rate representing population changes in number of infections polymers (P) is

$$\underbrace{\underbrace{b}_{\substack{\text{Rate of new chains created} \\ \text{(polymers breaking)}}}^{\text{Rate of new chains created}} \underbrace{\underbrace{(\mu_P + \mu_I)}^{\text{I}}}_{\text{Average time spent in Z}}$$

Multiplying these together gives $R_0^{NG}(D,\mu_I) = \frac{b(\beta S^* - n(n-1)b)}{(\mu_P + \mu_I)((\mu_P + \mu_I) + (n-1)b)}$. Thus, $R_0^{NG}(D,\mu_I)$ represents prion replication as a two stage process in which a prion must first grow longer and then break in order to create a new infectious chain. Thus, this value represents the secondary infections of P over two time steps, rather than one. As such, like many vector disease models, the geometric mean of $R_0^{NG}(D,\mu_I)$ is also a valid basic reproductive number. Interestingly enough, for the parameter values used in this work $R_0^{H}(D,\mu_I) \approx \sqrt{R_0^{NG}(D,\mu_I)}$

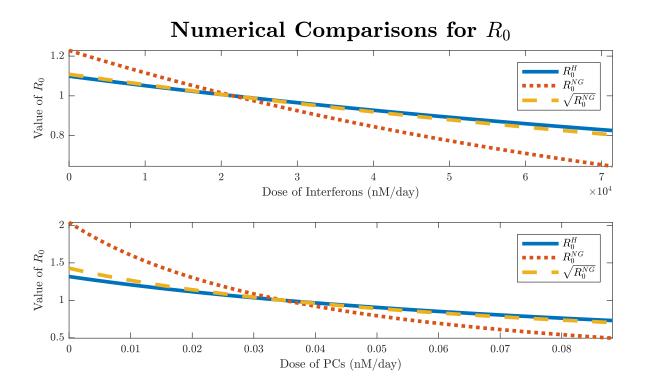


Figure 8: The geometric mean of $\mathbf{R}_0^{NG}(D, \mu_I)$ is approximately equal to the heuristic value, $\mathbf{R}_0^{H}(D, \mu_I)$

as can be seen in Figure 8.

When the treatment methods are being implemented, it can be seen that the interferon treatment lowers the R_0 by increasing the death terms of prion polymers and monomers from μ_P to $(\mu_P + \mu_I)$. The effect of the pharmacological chaperone treatment is not as immediately obvious. When the pharmacological chaperone dose is set to zero (D = 0), as in Masel et al. 1999, $S^* = \frac{\Lambda}{\mu_S}$. This means that when doses of pharmacological chaperones are being introduced, S^* is a different value. This S^* is less than $\frac{\Lambda}{\mu_S}$ because $S^* = \frac{\Lambda}{\mu_S} - T^* > 0$. Therefore, the pharmacological chaperone treatment lowers R_0 by lowering the number of susceptible PrP^C proteins at the PFE.

3.4 Endemic Equilibrium

To analyze the endemic equilibrium, assume that the interferons do not affect R, the noninfectious PrP^{Sc} monomers that are not joined to polymer chains (i.e. $\sigma = 0$). Due to the complexity of the system, the endemic equilibrium will only be analyzed in the special case where the antibody treatment is not being administered (D = 0). Unlike the prion-free equilibrium and secondary infection rate, the endemic equilibrium of System C.2 is different from that of Masel et al. (1999) when neither treatment is being administered (D = 0 and $\mu_I = 0$). When D = 0 the endemic equilibrium (EE) is given by $(S_E^*, T_E^*, P_E^*, Z_E^*, R_E^*, A_E^*)$ where:

$$S_{E}^{*} = \frac{\Lambda(\mu_{I} + \mu_{P})}{\mu_{S}} \frac{1}{R_{0}}$$

$$T_{E}^{*} = 0$$

$$P_{E}^{*} = \frac{\mu_{S}}{\beta_{S}}(R_{0} - 1)$$

$$Z_{E}^{*} = \frac{\mu_{S}((2n - 1)b + \mu_{I} + \mu_{P})}{b\beta_{S}}(R_{0} - 1)$$

$$R_{E}^{*} = \frac{n(n - 1)b}{\beta_{S}}(1 - R_{0}^{-1})$$

$$A_{E}^{*} = 0$$
(3.6)

Thus the endemic equilibrium exists only if $R_0 > 1$. Additionally, when $\mu_I = 0$, $S_E^* + R_E^*$ is equal to the value of S at the endemic equilibrium in Masel et al. 1999. This occurs because in Masel et al.'s paper, $\Pr P^{S_c}$ monomers are added back into the susceptible population, whereas in this paper, $\Pr P^{S_c}$ monomers are placed in the treatment resistant population (R).

3.5 Interferon Analysis

There are several experimental treatments that aim to stop prion proliferation, one of which is the direct dosing of interferons. Experiments done by Ishibashi et al. (2019) have indicated that the interferon signalling interferes with prion propagation. In this research, a group of mice was infected with prions and then were treated with different concentrations of interferons. The concentrations of prions was measured after 48 hours of the inoculation of interferons. Due to the interferon signalling, there was a decrease in the prion population. From this experiment, the following experimental data was obtained:

 $\{ \text{concentration (nM), ratio } P_I/P_0 \} = \{ 0.00001, 1.00 \}, \\ \{ 178970, 1.05 \}, \{ 894849, 0.93 \}, \{ 1.7897^*10^6, 0.945 \}, \{ 3.5794^*10^6, 0.90 \}, \{ 8.94849^*10^6, 0.65 \} \} \} \}$

For each of the samples treated with different concentrations, it is measured the ratio $P_{ratio} = P_I/P_0$ between the amount of prions before and after the interferons were introduced. Then, the data was interpolated with a Hill equation as shown in Figure 9.

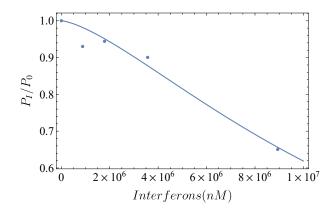


Figure 9: P_I/P_0 vs concentration. The ratio of prions measured in experiments with respect to the different concentrations of interferons. A Hill equation was used to interpolate the data.

From this interpolation the following equation was obtained.

$$P_{ratio} = \frac{1}{1 + \left(\frac{I}{13989.5 \times 10^3}\right)^{1.44313}} \tag{3.7}$$

where I is the concentration of interferons. Then, assuming that the dosage D and the death rate caused by interferons μ_I are constant, we rewrite the total death rate of the polymer chain $\mu_P + \mu_I$ as $\mu_P + (k)\mu_P$. We assume that the behavior of prions with respect to time (including the treatment) is given by an exponential decay:

$$P(t, I_0) = c_0 e^{-(1+k)\mu_p t}$$
(3.8)

The ratio between the amount of prions when there is treatment and when there is not treatment is given by

$$P_{\text{ratio}}(I) = \frac{P_I}{P_0} = \frac{c_0 e^{-(1+k)\mu_P t}}{c_0 e^{-\mu_P t}} = \frac{e^{-\mu_P t} e^{-ku_P t}}{e^{-\mu_P t}} = e^{-ku_P t}$$
(3.9)

Because the experiment dosed interferons over 48 hours, we assume t = 2 days and solve for k.

$$k(I) = -\frac{\log(P_{\text{ratio}}(I))}{2\mu_P} \tag{3.10}$$

where $P_{\text{ratio}}(I)$ is equation 3.5, a ratio between 0 and 1.

Note that the experimental data I was a unique dose. The P_{ratio} values were measured after two days of inoculation. It is assumed for calculations that the dose per day is the half of the total dose.

3.6 Growth Rate

The rate of exponential growth, r, is defined as the per capita change in number of chains of prions per unit of time [37]. According to Masel et al. (2000) [17], this parameter is the most important one to analyze the replication of infectious prions. It depends on the kinetics of the entire system and includes polymer replication and elongation. This exponential growth is the result of chains breaking into two polymers that are able to replicate. It is affected by changes in D (dosage of pharmacological chaperones) and μ_I (the interferon-induced prion death rate) and is therefore useful to analyze their impact on the system.

To calculate the growth rate, it is assumed that S, A, R, and T are initially at a steady state. This reduces our original closed system of differential equations to a linear system of equations for Z and P:

$$\frac{dP}{dt} = -(\mu_P + \mu_I)P + bZ - (2n - 1)bP,
\frac{dZ}{dt} = \beta SP + \beta RP - (\mu_P + \mu_I)Z - n(n - 1)bP,$$
(3.11)

Note that the relation between P and Z is linear at all times. In fact, Z = GP, where G

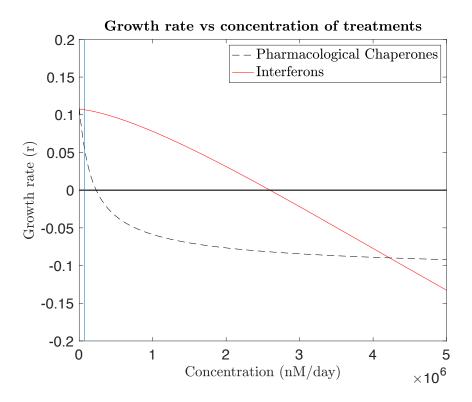


Figure 10: These plots shows the value of the prion growth rate against the concentrations of the pharmacological chaperone and interferon dosages.

is the average chain length (see 3.3). The rate r then determines the exponential growth of both P and Z. Then taking the Jacobian matrix and calculating the dominant eigenvalue of that matrix will provide the growth rate r. The system of P and Z will decay and accumulate exponentially at this rate r. The growth rate is given by

$$r = b(1-n) - \frac{b}{2} + \sqrt{\frac{b^2}{4} + \beta b(R^* + S^*)} - (\mu_I + \mu_p), \qquad (3.12)$$

where S^* and R^* represent the S and R values of the PFE. By replacing the value of k obtained in 3.10 into r, we obtain an expression that depends on the concentration of interferons. The exact derivation of the formula can be found in Appendix C.

Therefore, as the dosage of either treatment increases the growth rate decreases. Figure 10 particularly shows how I and D, the proposed treatments, directly affect the population of prions. To examine this system, r is plotted against our two treatment variables: μ_I , in order to study the effect of the interferon treatment, and D, which shows how the pharmacological

chaperones affect the growth rate of PrP^{Sc}, see Figure 10.

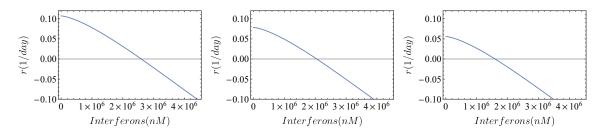


Figure 11: These plots shows the value of the prion growth rate against the concentrations of interferon dosages. From left to right, it is plotted for 3 different dosages: 0 nM, 500 000/4 nM, and 500 000/2 nM.

In Figure 11, holding D constant and plotting r against I, it can be seen that the growth rate decreases as I increases. Since the interferons remove $\Pr P^{Sc}$ from the population, this limits how fast P can grow. Figure 11 shows three plots for three different D, that indicate how the system is affected with the variation of D. This makes sense biologically because as pharmacological chaperones are added to the system, there are fewer $\Pr P^{C}$ monomers that can be added to $\Pr P^{Sc}$ chains. That means that P, the polymers, must grow more slowly when the dosage is increased. However, the pharmacological chaperones change the growth rate much less than the interferons do, and it becomes evident that I reduces the growth rate much more than D does. This indicates that the interferon treatment reduces the prion population much more than pharmacological chaperones do.

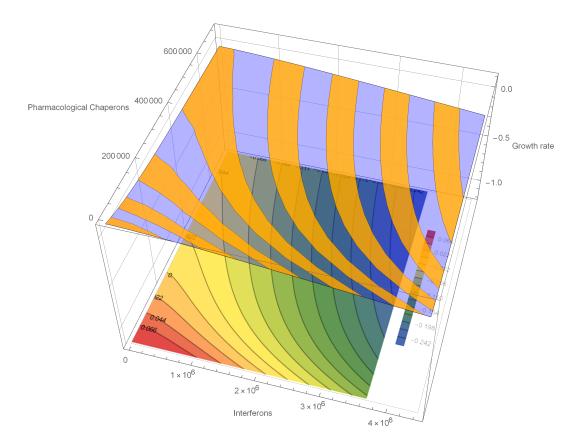


Figure 12: 3D representation of the value of the prion growth rate against the concentrations of the pharmacological chaperone and interferon dosages.

It was expected that the introduction of these treatments would reduce the concentration of prions in the brain. Our model shows that PrP^{Sc} decreases until it reaches the PFE by adding the treatment. Moreover, the model demonstrates the comparative efficacy of the treatments. From the analysis of the growth rate, we can see the first treatment has less efficacy. That is, more pharmacological chaperones are required to produce the desired result, which is to reduce the growth rate. Pharmacological chaperones, however, are more potent; that is, they reach their full effect more quickly. The interferon treatment is the opposite: more efficient but less potent. It can be seen that in Figure 14; R₀ decreases with both treatments, increasing the pharmacological chaperones or increasing the interferons. The range for both are different; big changes in pharmacological chaperones make no significant changes in R₀, and low changes in interferons make considerable changes in R₀. Similar changes are evident in r, the growth rate of the prion population. Increasing the pharmacological chaperone dose decreases r, but only by a small amount. On the other hand the

Name	Value	Units	Reference
Initial Conditions			
S	600	nM	[9]
T	0	nM	
P	3	nM	
Z	90	nM	
R	0	nM	
A	0	nM	
Parameters			
Λ	2400	$\frac{1}{\frac{\mathrm{day}}{1}}$	[31]
μ_S	4	$\frac{1}{dav}$	[31]
μ_P	0.047	$\frac{\overline{\mathrm{day}}}{\frac{1}{\mathrm{day}}}$	[31]
μ_R	4	$\frac{1}{\text{dav}}$	[31]
μ_A	62.0352	$\frac{1}{\text{dav}}$	[19]
μ_I	[0, 0.0882]	$\frac{1}{day}$ $\frac{1}{day}$ $\frac{1}{day}$	
σ	1	_	
β_S	0.00292	$\frac{1}{(\text{day})\text{nM}}$	[31]
β_R	0.00292	$\frac{\frac{1}{(\text{day})nM}}{\frac{1}{(\text{day})nM}}$	[31]
lpha	0.051408	$\frac{\frac{1}{1}}{(\text{day})\text{nM}}$	[19]
$b_{i,j}$	0.0314	$\frac{1}{\text{day}}$	[31]
n	3	—	[31, 33]
D	[0,71500]	$\frac{\mathrm{nM}}{\mathrm{day}}$	[15]

Table 3: Table of parameters for the numerical simulations. nM = nano-moles/Liter

introduction of interferons lowers the growth rate significantly. Just like with R_0 , big changes in pharmacological chaperones have a small effect on r while small doses of interferons have a large impact on r.

4 Numerical Analysis

This section includes the numerical simulations used to study the effect of treatment on prion proliferation in the brain. Parameters values were obtained from literature, particularly from Rubeinstein et al. [31]. For the parameters values see Table 3.

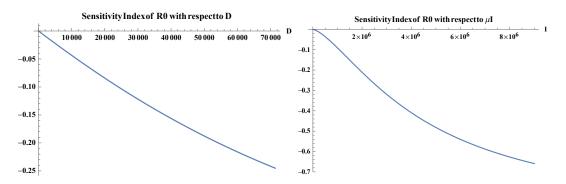


Figure 13: Sensitivity indexes of R_0 . Left: with respect to D against different values of D. Right: with respect to μ_I against different values of I.

4.1 Sensitivity Analysis

Sensitivity analysis was done on the basic reproductive number with respect to the treatmentrelated parameter values, D and μ_I . The sensitivity indexes were found by taking partial derivatives of R_0 with respect to either D or μ_I and then normalizing these values by multiplying them by $\frac{D}{R_0}$ or $\frac{\mu_I}{R_0}$, respectively. The value of the sensitivity index of R_0 with respect to D represents the percent change in R_0 when D is increased by 1%. In Figure 13, it can be seen that within the parameter ranges under consideration in this work, R_0 is more sensitive to interferon treatment than pharmacological chaperone treatment, which indicates that the pharmacological chaperone treatment will be more effective at reducing the growth rate of the prion population initially.

4.2 Numerical Results

The model presented in this paper concerns the effectiveness of two treatments, pharmacological chaperones and interferons, so this analysis examines which of these treatments affects the population of PrP^{Sc} the most.

Figure 14 compares R_0 values at time t_{final} with respect to different values of D and μ_I , respectively. From this graph, it is evident that the more each treatment increases, the more R_0 lowers, indicating the treatments are effective. Note that R_0 decreases much more with respect to an increase in μ_I , the death rate induced by interferons. This suggests that using interferons is the more effective treatment for prion diseases. It is important to note that

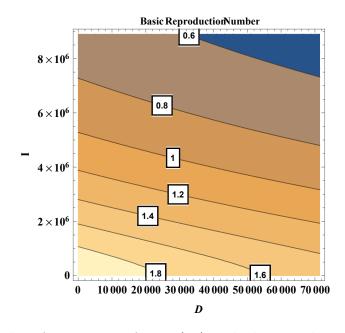


Figure 14: Contour plot of R_0 vs. Interferons (μ_I) and Pharmacological Chaperone Dosage (D)

the pharmacological chaperones affect R_0 as well. However, these results indicate that the secondary infection rate is much more sensitive to the changes in μ_I , a sign that treatment of prion diseases with interferons may be more useful.

From Figure 15, the relationship between the endemic value of P compared to the treatment levels is evident. When no treatment is applied, P_E is at its highest in the lower left corner of the figure. This makes sense; treatments should limit the strength of the final infection level. Travelling along the I axis, it is observed that P_E decreases steeply the more the interferon treatment increases. This makes sense with the previous analysis of r, the growth rate; the interferon treatment is the more effective of the two treatments. The decrease that happens along the D-axis is much less pronounced, but still existent. Note that the pharmacological chaperone dosage is limited due to toxicity of the treatment. Pharmacological chaperones affect the amount of PrP^{Sc} as well, though not as much the interferons do.

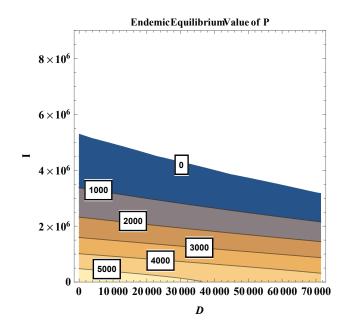


Figure 15: Contour plot of P_E vs. Interferons (μ_I) and Dosage (D)

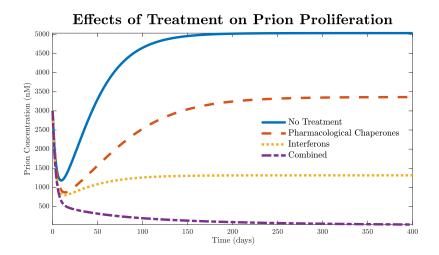


Figure 16: This figure shows a special case of Figure 14. Prion concentration changes over time with various treatments implemented. Pharmacological chaperons and interferons are represented at , $7.1429 * 10^4$ nM/day, and $1.25 * 10^6$ nM/day respectively. Each of these dosages alone will go to endemic equilibrium. This figure shows that the treatments are able to be combined at the same dosages to go to disease free equilibrium. With no treatment, the prion concentration increase faster than with treatments. (For parameters, see Table 3).

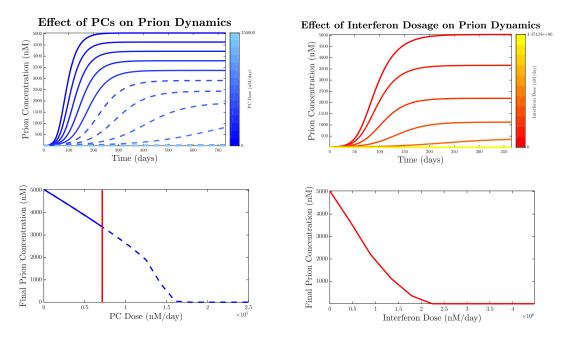


Figure 17: Interferon Dosages vs. Time. **Top:** The concentration of prions as a function of time, where each line is a different dose of pharmacological chaperones (left) or interferons (right). **Bottom:** The final prion concentration as a function of dose of pharmacological (left) or interferons (right). Dotted lines on the leftmost graphs are doses beyond the toxic threshold of pharmacological chaperones. The red line on the bottom left graph represents the toxic dose of pharmacological chaperones. (For parameter values, see Table 3)

Figure 16 effectively shows the synergistic effect of the two treatment's on the endemic equilibrium (P_E) . While interferons are more effective than pharmacological chaperones at both reducing the value of the endemic equilibrium and the time it takes to reach that equilibrium, the combined treatment shows a near 100-fold decrease in endemic equilibrium than that of the untreated infection.

Figure 17 shows the differing concentration of prions at a range of constant interferon dosages. It can be shown that at these particular initial conditions found in Table 3, there is a transition of stable equilibrium from a stable endemic equilibrium at low dosages of interferons to a stable prion-free equilibrium at higher dosages. As the interferon dose increases, the prion concentration does not increase as quickly as without treatment. The latter two graphs in the figure shows how the final concentration of prions depends on interferons; this value goes to zero as we increase both interferon dose and, relatedly, the interferon-induced death rate. This treatment, at high enough doses, appears to work against prion proliferation. The prion concentration over time with different constant dosages of pharmacological chaperones and no interferon treatment was also examined. Figure 16 shows how combination of lower-dose treatments can be effective. The dosages administered of both pharmacological chaperones and interferons alone cannot bring the prion population to zero. Each lowers the endemic equilibrium but does not eliminate the prion population. However, if the low doses of both treatments are combined, they can work together to bring the prion concentration down together. The synergistic effect is significant.

Figure 18 shows the combination of interferon and pharmacological chaperone treatments. The red and orange lines are the same as in Figure 17. The superimposed blue and purple lines show the effect of introducing a constant dose of pharmacological chaperones while varying interferon dose. The upper limit for the prion concentration is much lower than with a single treatment. The dual treatment also slows the growth of PrP^{Sc} more than interferons alone does. The second and third graphs in Figure 18 show how interferons affect the final prion concentration, again with both no pharmacological chaperones (red) and a constant dose of pharmacological chaperone treatment (blue). The addition of a constant treatment dose means that the final prion concentration is achieved sooner. This graph shows us that using the two treatments is a way to reduce prion concentration more and faster than either pharmacological chaperones or interferons alone.

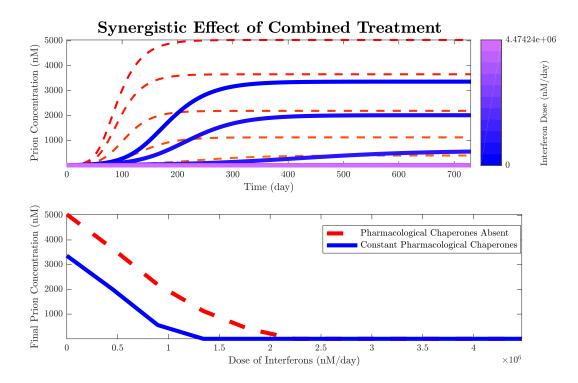


Figure 18: Combined Interferon and Pharmacological Chaperones. **Top:** The concentration of prions as a function of interferon and pharmacological chaperone doses. Red is only pharmacological chaperone doses, while blue represents varying interferon doses and constant pharmacological chaperones. Pharmacological chaperons are at a constant dose of $7.1429 * 10^4$, which is right at the toxic threshold. **Bottom:** The final prion concentration as a function of dose of interferons, with and without the same doses of pharmacological chaperones as above.

5 Conclusions

In this paper, previous models by Nowak et al. [23] and Masel et al. [18] on prion dynamics in the brain were examined. Our work introduces two possible treatments for prion diseases. The first treatment uses pharmacological chaperones, which prevent PrP from misfolding into PrP^{Sc}. The other treatment uses interferons, a signalling part of the immune system that reduces the PrP^{Sc} population. This work examined how these treatments affected the population of prions in the brain.

From the results of the numeric simulations it was clear that both treatments affect prion proliferation. However, through analysis of the basic reproduction number (R_0) and the prion growth rate (r), it can be seen how the treatments work in tandem. The pharmacological chaperones act quickly in reducing the PrP^{Sc} population, but their effect does not last as long. On the contrary, interferon treatment does not work as quickly, but it works more effectively as time goes on. The best treatment is a combination of pharmacological chaperones and interferons. The pharmacological chaperones act more quickly but with less potency, meaning that they are a good first action to take with prion diseases. Interferons are the more potent treatment, meaning that they reduce the prion population far more than pharmacological chaperones do over time. However, this effect takes longer to occur. Thus, in this work it is shown that the best treatment is likely a combination of the two. Pharmacological chaperones act quickly, reducing the amount of prions, buying some time for the interferon treatment to work. This might suggest that pharmacological chaperones should be administered before the interferons.

It is important to note that this is not a cure. Interferons and pharmacological chaperones have only been shown to prolong the life of someone afflicted with a prion disease, not eliminate the disease altogether. Further research beyond this work is needed if any steps are to be made towards a cure. Mathematically, the optimization of the two treatments would be useful. Looking for specific dosages of both pharmacological chaperones and interferons and best timing of these doses is also important. The results in this paper would be best supported by a toxicity optimization; like all drugs, the treatments presented here can not be given to a patient freely. What would be the best combination of treatment to reduce harm done by the drugs while still affecting the prion population? Additionally, further research of other possible treatments and treatment combinations is essential. Pharmacological chaperones and interferons are not the only two possible ways to treat prion diseases, and it is important to take into account all possible therapies. In vivo experiments are necessary to show if these treatments actually have any affect on prion diseases. Once more substantial research has been done, a cost analysis of these treatments would also be useful to reduce cost for the patient. Prion diseases are still fatal, still dangerous, still incurable. But research is happening, and perhaps we are one step closer to solving the mystery of these strange diseases.

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References

- A. Andreadis, Generation Of Protein Isoform Diversity By Alternative Splicing: Mechanistic And Biological Implications, Annual Review of Cell and Developmental Biology 3 (1987), no. 1, 207–242.
- [2] S. V. Antonyuk, C. R. Trevitt, R. W. Strange, G. S. Jackson, D. Sangar, M. Batchelor, S. Cooper, C. Fraser, S. Jones, T. Georgiou, A. Khalili-Shirazi, A. R. Clarke, S. S. Hasnain, and J. Collinge, *Crystal structure of human prion protein bound to a therapeutic antibody*, Proceedings of the National Academy of Sciences **106** (2009), no. 8, 2554–2558.
- [3] B. Beutler, Z. Jiang, P. Georgel, K. Crozat, B. Croker, S. Rutschmann, X. Du, and K. Hoebe, GENETIC ANALYSIS OF HOST RESISTANCE: Toll-Like Receptor Signaling and Immunity at Large, Annual Review of Immunology 24 (2006), no. 1, 353–389.

- [4] S. R. Collins, A. Douglass, R. D. Vale, and J. S. Weissman, Mechanism of Prion Propagation: Amyloid Growth Occurs by Monomer Addition, PLoS Biology 2 (2004), no. 10, 1582–1590.
- [5] M. De Andrea, R. Ravera, D. Gioia, M. Gariglio, and S. Landolfo, *The interferon system: An overview*, European Journal of Paediatric Neurology 6 (2002), no. Suppl. A, A42–A46.
- [6] Center for Disease Control and Prevention (CDC), Prion Diseases", https://www.cdc.gov/prions/index.html.
- [7] Center for Disease Control and Prevention, Ocurrence and Transmission Creutzfeldt-Jakbo Disease Classic (CJD), https://www.cdc.gov/prions/cjd/ occurrence-transmission.html.
- [8] M. L. Greer, L. Pujo-Menjouet, and G. F. Webb, A mathematical analysis of the dynamics of prion proliferation, Journal of Theoretical Biology 242 (2006), no. 3, 598–606.
- [9] L. Gregori, B. N. Gray, E. Rose, D. S. Spinner, R. J. Kascsak, and R. G. Rohwer, A sensitive and quantitative assay for normal PrP in plasma, Journal of Virological Methods 149 (2008), no. 2, 251 – 259.
- [10] E. C. Gunther, L. M. Smith, M. A. Kostylev, T. O. Cox, A. C. Kaufman, S. Lee, E. Folta-Stogniew, G. D. Maynard, J. W. Um, M. Stagi, J. K. Heiss, A. Stoner, G. P. Noble, H. Takahashi, L. T. Haas, J. S. Schneekloth, J. Merkel, C. Teran, Z. K. Naderi, S. Supattapone, and S. M. Strittmatter, *Rescue of Transgenic Alzheimer's Pathophysi*ology by Polymeric Cellular Prion Protein Antagonists, Cell Reports 26 (2019), no. 1, 145–158.e8.
- [11] D. Ishibashi, R. Atarashi, T. Fuse, T. Nakagaki, N. Yamaguchi, K. Satoh, K. Honda, and N. Nishida, Protective role of interferon regulatory factor 3-mediated signaling against prion infection, Journal of virology 86 (2012), no. 9, 4947–4955.

- [12] D. Ishibashi, T. Homma, T. Nakagaki, T. Fuse, K. Sano, K. Satoh, T. Mori, R. Atarashi, and N. Nishida, *Type I interferon protects neurons from prions in in vivo models*, Brain: a journal of neurology **142** (2019), no. 4, 1035–1050.
- [13] Y. Iturria-Medina, R. C. Sotero, P. J. Toussaint, and A. C. Evans, Epidemic Spreading Model to Characterize Misfolded Proteins Propagation in Aging and Associated Neurodegenerative Disorders, PLoS Computational Biology 10 (2014), no. 11, 1–15.
- [14] Y. O. Kamatari, Y. Hayano, K. Yamaguchi, J. Hosokawa-Muto, and K. Kuwata, Characterizing antiprion compounds based on their binding properties to prion proteins: Implications as medical chaperones, Protein Science 22 (2013), no. 1, 22–34.
- [15] D. A. Kocisko, W. S. Caughey, R. E. Race, G. Roper, B. Caughey, and J. D. Morrey, A Porphyrin Increases Survival Time of Mice after Intracerebral Prion Infection, Antimicrobial Agents and Chemotherapy 50 (2006), no. 2, 759–761.
- [16] J. Kubelka, J. Hofrichter, and W. A. Eaton, *The protein folding "speed limit*", Current Opinion in Structural Biology 14 (2004), no. 1, 76 – 88.
- [17] J. Masel and V. A.A. Jansen, Designing drugs to stop the formation of prion aggregates and other amyloids, Biophysical Chemistry 88 (2000), no. 1-3, 47–59.
- [18] J. Masel, V. AA Jansen, and M. A. Nowak, Quantifying the kinetic parameters of prion replication, Biophysical chemistry 77 (1999), no. 2-3, 139–152.
- [19] T. Massignan, S. Cimini, C. Stincardini, M. Cerovic, I. Vanni, S. R. Elezgarai, J. Moreno, M. Stravalaci, A. Negro, V. Sangiovanni, and et al., A cationic tetrapyrrole inhibits toxic activities of the cellular prion protein, Scientific Reports 6 (2016), no. 1.
- [20] D. L. Mobley, D. L. Cox, R. R. P. Singh, R. V. Kulkarni, and A. Slepoy, Simulations of Oligomeric Intermediates in Prion Diseases, Biophysical Journal 85 (2003), no. 4, 2213–2223.

- [21] J. Morello, M. Bouvier, U. E. Petäjä-Repo, and D. G. Bichet, *Pharmacological chaper-ones: A new twist on receptor folding*, Trends in pharmacological sciences **21** (2000), no. 12, 466–469.
- [22] A. Nicoll, C. R. Trevitt, M. H. Tattum, E. Risse, E. Quarterman, A. Ibarra, C. Wright, G. Jackson, R. Sessions, M. Farrow, J. P. Waltho, A. R. Clarke, and J. Collinge, *Phar*macological chaperone for the structured domain of human prion protein, Proc Natl Acad Sci U S A **107** (2010), 17610–17615.
- [23] M. A. Nowak, D. C. Krakauer, A. Klug, and R. M. May, Prion Infection Dynamics, Integrative Biology (1998), 3–15.
- [24] D. Peretz, R. Anthony Williamson, K. Kaneko, J. Vergara, E. Leclerc, G. Schmitt-Ulms, I. Mehlhorn, G. Legname, M. Wormald, P. Rudd, R. Dwek, D. Burton, and S. Prusiner, *Antibodies inhibit prion propagation and clear cell cultures of prion infectivity*, Nature 412 (2001), 739–743.
- [25] M. Polymenidou, R. Moos, M. Scott, C. Sigurdson, Y. Shi, B. Yajima, I. Hafner-Bratkovic, R. Jerala, S. Hornemann, K. Wuthrich, A. Bellon, M. Vey, G. Garen, M. James, N. Kav, and A. Aguzzi, *The POM Monoclonals: A Comprehensive Set of Antibodies to Non-Overlapping Prion Protein Epitopes*, PloS one **3** (2008), e3872.
- [26] T. Pöschel, N. V. Brilliantov, and C. Frömmel, *Kinetics of Prion Growth*, Biophysical Journal 85 (2003), no. 6, 3460–3474.
- [27] E. T. Powers and D. L. Powers, The kinetics of nucleated polymerizations at high concentrations: Amyloid fibril formation near and above the "supercritical concentration", Biophysical Journal 91 (2006), no. 1, 122–132.
- [28] S. B. Prusiner, *Prions*, Proceedings of the National Academy of Sciences of the United States of America 95 (1998), 13363–13383.

- [29] C. Riemer, I. Queck, D. Simon, R. Kurth, and M. Baier, Identification of Upregulated Genes in Scrapie-Infected Brain Tissue, Journal of Virology 74 (2000), no. 21, 10245–10248.
- [30] Wencel Valega-mackenzie Karen R Ríos-soto, Can Vaccination Save a Zika Virus Epidemic ?, Bulletin of Mathematical Biology (2018).
- [31] R. Rubenstein, P. C. Gray, T. Cleland, M. S. Piltch, W. Hlavacek, R. Roberts, J. Ambrosiano, and J-I Kim, *Dynamics of the nucleated polymerization model of prion replication*, Biophysical chemistry **125** (2007), 360–7.
- [32] A. Sagare, M. Sweeney, A. Nelson, Z. Zhao, and B. Zlokovic, Prion Protein Antagonists Rescue Alzheimer's Amyloid-β-Related Cognitive Deficits, Trends in Molecular Medicine 25 (2019), 74–76.
- [33] J. R. Silveira, G. J. Raymond, A. G. Hughson, R. E. Race, V. L. Sim, S. F. Hayes, and B. Caughey, *The most infectious prion protein particles*, 437 (2005), no. September, 257–261.
- [34] C. Soto and N. Satani, The intricate mechanisms of neurodegeneration in prion diseases, Trends in molecular medicine 17 (2011), no. 1, 14–24.
- [35] M. J. Stobart, D. Parchaliuk, S. L. R. Simon, J. LeMaistre, J. Lazar, R. Rubenstein, and J. D. Knox, Differential expression of interferon responsive genes in rodent models of transmissible spongiform encephalopathy disease, Molecular Neurodegeneration 2 (2007), no. 1, 5.
- [36] Prion Unit, Drug treatments, http://www.prion.ucl.ac.uk/clinic-services/ research/drug-treatments.
- [37] J. Wallinga and M. Lipsitch, How generation intervals shape the relationship between growth rates and reproductive numbers, Proceedings of the Royal Society B: Biological Sciences 274 (2007), no. 1609, 599–604.

- [38] T. Wisniewski, J. A. Chabalgoity, and F. Goni, Is vaccination against transmissible spongiform encephalopathy feasible?, Rev Sci Tech (2007).
- [39] C. Zhanhua, J. Gah-Kok Gan, L. Lei, M. Sakharkar, and P. Kangueane, Protein subunit interfaces: Heterodimers versus homodimers, Bioinformation 1 (2005), 28–39.

A Mathematical Proofs for the Equations

A.1 Derivation of the Infinite System of Differential Equations

In this section, the derivation of the system of differential equations from the infinite dimensional system is explained. First, take the term $\sum_{i=1}^{n-1} \sum_{j=i+1}^{\infty} (b_{j,i} + b_{i,i-j})iP_j$, which describes a polymer of length *i* splitting into two polymers, with one of them of length less than *n*. When the shorter polymer is below the *n* threshold, it dissociates into infected monomers, and so the monomers go from the P_i class to *R*. Thus, for a single length *j* we can represent the monomers flowing into *R* by the following:

$$b_{j,1}P_j + 2b_{j,2}P_j + \ldots + (n-1)b_{j,n-1}P_j + (n-1)b_{j,j-n+1}(n-1)P_j + \ldots + 2b_{j,j-2}P_j + b_{j,j-1}P_j,$$

The sum of this value over all lengths j can be written,

$$\sum_{j=1}^{\infty} (b_{j,1}P_j + \dots + (n-1)b_{j,n-1}P_j + (n-1)b_{j,j-n+1}(n-1)P_j + \dots + b_{j,j-1}P_j)$$
$$= \sum_{j=1}^{\infty} \sum_{i=1}^{n-1} i(b_{j,i}P_j + b_{j,j-i}P_j).$$
(A.1)

It is known that $b_{j,i} = 0$ if $i \ge j$; so Equation A.1 can be rewritten as:

$$\sum_{i=1}^{n-1} \sum_{j=i+1}^{\infty} i(b_{j,i}P_j + b_{j,j-i}P_j).$$

Different assumptions about the dynamics of aggregate growth and fragmentation can

be embodied in the matrix $b_{i,j}$. In this, we assume that $b_{i,j} = b$, which changes the sum as follows:

$$\sum_{j=1}^{i-1} b_{i,j} P_i = b P_i \sum_{j=1}^{i-1} 1 = b(i-1) P_i,$$
(A.2)

$$\sum_{j=i+1}^{\infty} (b_{j,i} + b_{i,i-j}) P_j = 2b \sum_{j=i+1}^{\infty} P_j, \text{ and}$$
(A.3)

$$\sum_{i=1}^{n-1} \sum_{j=i+1}^{\infty} (b_{j,i} + b_{i,i-j}) i P_j = 2b \sum_{i=1}^{n-1} \sum_{j=i+1}^{\infty} i P_j.$$
(A.4)

Therefore, the construction of the infinite dimensional system can be written as:

$$\begin{aligned} \frac{dS}{dt} &= \Lambda - \mu_S S - \alpha A S + \mu_A T - \beta_S S \sum_{i=n}^{\infty} P_i, \\ \frac{dT}{dt} &= -\mu_S T + \alpha A S - \mu_A T, \\ \frac{dP_i}{dt} &= -\mu_P P_i - \mu_I P_i + \beta_S S P_{i-1} - \beta_S S P_i + \beta_R R P_{i-1} - \beta_R R P_i - b(i-1) P_i + \\ & 2b \sum_{j=i+1}^{\infty} P_j, \\ \frac{dR}{dt} &= -\mu_R R - \mu_I(t) \sigma R - \beta_R R \sum_{i=n}^{\infty} P_i + 2b \sum_{i=1}^{n-1} \sum_{j=i+1}^{\infty} i P_j, \text{ and} \\ \frac{dA}{dt} &= -\mu_A A - \alpha A S + D, \end{aligned}$$

as it is expressed in System 2.1.

A.2 Closing the System of Differential Equations

Here we close the system with infinite equations by summing over P_i . Let

$$P = \sum_{i=n}^{\infty} P_i, \quad Z = \sum_{i=n}^{\infty} i P_i.$$
(A.5)

Therefore, it is assumed that these sums are convergent since in any biological system, there will only be a finite number of prions. Their derivatives can be written as $\dot{P} = \sum_{i=n}^{\infty} \dot{P}_i$ and

 $\dot{Z} = \sum_{i=n}^{\infty} i \dot{P}_i$. The derivative of P can then be reduced as follows:

$$\dot{P} = \sum_{i=n}^{\infty} \left(\beta SP_{i-1} - \beta SP_i + \beta RP_{i-1} - \beta RP_i - (\mu_p + \mu_I)P_i - b(i-1)P_i + 2b \sum_{j=i+1}^{\infty} P_j \right)$$

$$=\underbrace{\sum_{i=n}^{\infty}\beta S(P_{i-1}-P_i)}_{\text{Telescopic sum}} +\underbrace{\sum_{i=n}^{\infty}\beta R(P_{i-1}-P_i)}_{\text{Telescopic sum}} -\sum_{i=n}^{\infty}(\mu_p+\mu_I)P_i - \sum_{i=n}^{\infty}b(i-1)P_i + \sum_{i=n}^{\infty}2b\sum_{j=i+1}^{\infty}P_j$$

 $= -(\mu_p + \mu_I) \sum_{\substack{i=n \\ P}}^{\infty} P_i - b \sum_{\substack{i=n \\ P}}^{\infty} (i-1)P_i + 2b \sum_{\substack{i=n \\ I=n}}^{\infty} \sum_{j=i+1}^{\infty} P_j$ $= -(\mu_p + \mu_I)P - b \sum_{\substack{i=n \\ Z}}^{\infty} iP_i + b \sum_{\substack{i=n \\ P}}^{\infty} P_i + 2b \sum_{\substack{i=n \\ I=n}}^{\infty} (i-n)P_i - 2b(n-n)P_n$ $= -(\mu_p + \mu_I)P - bZ + bP + 2b \sum_{\substack{i=n \\ Z}}^{\infty} iP_i - 2bn \sum_{\substack{i=n \\ P}}^{\infty} P_i.$

Thus,

$$\dot{P} = -(\mu_p + \mu_I)P - bZ + bP + 2bZ - 2bnP = -(\mu_p + \mu_I)P + bZ - (2n-1)bP.$$
(A.6)

The derivative of Z can also be rewritten:

$$\dot{Z} = \sum_{i=n}^{\infty} i \left(\beta SP_{i-1} - \beta SP_i + \beta RP_{i-1} - \beta RP_i - (\mu_p + \mu_I)P_i - b(i-1)P_i + 2b \sum_{j=i+1}^{\infty} P_j \right)$$
$$= \sum_{i=n}^{\infty} i \beta S(P_{i-1} - P_i) + \sum_{i=n}^{\infty} i \beta R(P_{i-1} - P_i)$$
$$- \sum_{i=n}^{\infty} i (\mu_p + \mu_I)P_i - \sum_{i=n}^{\infty} i b(i-1)P_i + \sum_{i=n}^{\infty} i 2b \sum_{j=i+1}^{\infty} P_j.$$
(A.7)

The sums in Equation A.7 can be reduced as follows:

$$\beta S \sum_{i=n}^{\infty} i(P_{i-1} - P_i) = \beta S(n(P_{n-1} - P_n) + (n+1)(P_n - P_{n+1}) + (n+2)(P_{n+1} - P_{n+2}) + \dots)$$

$$=\beta S(nP_{n-1} + P_n(n+1-n) + P_{n+1}(n+2-n-1) + P_{n+2}(n+3-n-2) + \dots)$$
$$=\beta S(P_n + P_{n+1} + P_{n+2} + P_{n+3} + P_{n+4} + \dots) = \beta S\sum_{i=n}^{\infty} P_i = \beta SP,$$

$$\beta R \sum_{i=n}^{\infty} i(P_{i-1} - P_i) = \beta R(n(P_{n-1} - P_n) + (n+1)(P_n - P_{n+1}) + (n+2)(P_{n+1} - P_{n+2}) + \dots)$$
$$= \beta R(nP_{n-1} + P_n(n+1-n) + P_{n+1}(n+2-n-1) + P_{n+2}(n+3-n-2) + \dots)$$
$$= \beta R(P_n + P_{n+1} + P_{n+2} + P_{n+3} + P_{n+4} + \dots) = \beta R \sum_{i=n}^{\infty} P_i = \beta RP, \text{ and}$$

$$\begin{aligned} 2b\sum_{i=n}^{\infty} i\sum_{j=i+1}^{\infty} P_j &= 2b\left(n\sum_{j=n+1}^{\infty} P_j + (n+1)\sum_{j=n+2}^{\infty} P_j + (n+2)\sum_{j=n+3}^{\infty} P_j + \ldots\right) \\ &= 2b\left(nP_{n+1} + (n+(n+1))P_{n+2} + (n+(n+1) + (n+2))P_{n+3} + \ldots\right) \\ &= 2b\left(\sum_{i=n+1}^{\infty} \frac{(i-1-n)(i-1+n) + i - 1 + n}{2}P_i\right) \\ &= 2b\left(\sum_{i=n+1}^{\infty} \frac{(i-1+n)((i-1-n) + 1)}{2}P_i\right) = 2b\left(\sum_{i=n+1}^{\infty} \frac{(i-1+n)(i-n)}{2}P_i\right) \\ &= 2b\left(\sum_{i=n}^{\infty} \frac{(i-1+n)(i-n)}{2}P_i\right) = 2b\left(\sum_{i=n}^{\infty} \frac{(i-1+n)(n-n)}{2}P_i\right) \\ &= 2b\left(\sum_{i=n}^{\infty} \frac{(i-1+n)(i-n)}{2}P_i\right) = 2b\left(\sum_{i=n}^{\infty} \frac{(i-1)i - n(i-1) + ni - n^2}{2}P_i\right) \\ &= 2b\left(\sum_{i=n}^{\infty} \frac{i^2 - i - ni + n + ni - n^2}{2}P_i\right) = 2b\left(\sum_{i=n}^{\infty} \frac{i^2 - i + n - n^2}{2}P_i\right) \\ &= 2b\sum_{i=n}^{\infty} \frac{i(i-1) - n(n-1)}{2}P_i = b\sum_{i=n}^{\infty} i(i-1)P_i - b\sum_{i=n}^{\infty} n(n-1)P_i. \end{aligned}$$

Plugging these reduced terms back into \dot{Z} yields

$$\dot{Z} = \sum_{i=n}^{\infty} i\beta S(P_{i-1} - P_i) + \sum_{i=n}^{\infty} i\beta R(P_{i-1} - P_i) - (\mu_p + \mu_I) \sum_{i=n}^{\infty} iP_i - b \sum_{i=n}^{\infty} i(i-1)P_i + \sum_{i=n}^{\infty} i2b \sum_{j=i+1}^{\infty} P_j$$

$$=\beta SP + \beta RP + (\mu_p + \mu_I)Z - b\sum_{i=n}^{\infty} i(i-1)P_i + b\sum_{i=n}^{\infty} i(i-1)P_i - b\sum_{i=n}^{\infty} n(n-1)P_i$$
$$=\beta SP + \beta RP + (\mu_p + \mu_I)Z - bn(n-1)\sum_{i=n}^{\infty} P_i.$$

Thus,

$$\dot{Z} = \beta SP + \beta RP + (\mu_p + \mu_I)Z - n(n-1)bP.$$
(A.8)

Now the infinite sums in Equation 2.2e (shown here in Equation A.9) can be reduced.

$$\dot{R} = 2b \sum_{i=1}^{n-1} \sum_{j=i+1}^{\infty} iP_j - \beta_R R \sum_{i=n}^{\infty} P_j - \mu_0 R.$$
(A.9)

These sums can be reduced as follows:

$$2b\sum_{i=1}^{n-1}\sum_{j=i+1}^{\infty}iP_j = 2b\left(\sum_{j=2}^{\infty}P_j + 2\sum_{j=3}^{\infty}P_j + 3\sum_{j=4}^{\infty}P_j + \dots + (n-1)\sum_{j=n}^{\infty}P_j\right)$$
$$= 2b\left(\sum_{i=1}^{n-1}\frac{i(i+1)}{2}P_i + \sum_{i=n}^{\infty}\frac{n(n-1)}{2}P_i\right)$$
$$= 2b\sum_{i=n}^{\infty}\frac{n(n-1)}{2}P_i = bn(n-1)\sum_{i=n}^{\infty}P_i = bn(n-1)P,$$

because $P_i = 0$ for i < n. Thus,

$$\dot{R} = -\mu_0 R - \beta RP + n(n-1)bP.$$

A.3 Stability Condition of the Prion Free Equilibrium

The linearization matrix of System 2.2 around $E^* = (S^*, T^*, 0, 0, 0, A^*)$ is

$$\mathbf{J}(\mathbf{E}^*) =$$

$$\begin{pmatrix} -\mu_S - \alpha A^* & \mu_A & -\beta_S S^* & 0 & 0 & -\alpha S^* \\ \alpha A^* & -\mu_S - \mu_A & 0 & 0 & 0 & \alpha S^* \\ 0 & 0 & -\mu_P - \mu_I - (2n-1)b & b & 0 & 0 \\ 0 & 0 & \beta_S S^* - n(n-1)b & -\mu_P - \mu_I & 0 & 0 \\ 0 & 0 & n(n-1)b & 0 & -\mu_R - \mu_I \sigma & 0 \\ -\alpha A^* & 0 & 0 & 0 & 0 & -\mu_A - \alpha S^* \end{pmatrix}$$

The eigenvalues of the matrix $J(E^*)$ are:

an

$$\begin{split} \lambda_1 &= -\mu_S < 0, \\ \lambda_2 &= -(\mu_R + \mu_I \sigma) < 0, \\ \lambda_3 &= -\frac{1}{2}b(2n-1) - (\mu_I + \mu_P) + \frac{1}{2}\sqrt{4S^*b\beta_S + b^2}, \\ \lambda_4 &= -\frac{1}{2}b(2n-1) - (\mu_I + \mu_P) - \frac{1}{2}\sqrt{4S^*b\beta_S + b^2} < 0, \\ \lambda_5 &= -\frac{\alpha}{2}(A^* + S^*) - \mu_A - \frac{1}{2}\mu_S + \frac{1}{2}\sqrt{M}, \\ d \quad \lambda_6 &= -\frac{\alpha}{2}(A^* + S^*) - \mu_A - \frac{1}{2}\mu_S - \frac{1}{2}\sqrt{M}, \end{split}$$

where $M = \alpha^2 (A^* + S^*)^2 + 2\alpha \mu_S (A^* - S^*) + \mu_S^2 > 0$, if $\phi_2 = (\mu_A + \mu_S)(\Lambda \alpha - \mu_S \sqrt{\mu_A^2 + 4\alpha D}) + \sqrt{\Delta} - \alpha \mu_S D < 0$.

(i)From the value of λ_3 , we have the $\lambda_3 < 0$ when $\phi_1 < 0$, where $\phi_1 = \sqrt{\Delta} + (\Lambda \alpha - \mu_A \mu_S)(\mu_A + \mu_S) - \alpha \mu_S D$.

(ii)From the value of λ_4 , we have the $\lambda_4 < 0$ if $n > \frac{1}{2}$.

(iii)From the value of λ_5 , because $4\mu_A(\mu_S + \mu_A) + 4\alpha(\mu_S S^* + \mu_A S^* + \mu_A A^*) > 0$, so $\lambda_5 < 0$. But, we want M is positive. Hence, we need $\phi_2 < 0$.

As a result, equilibrium E^* is locally stable when $\phi_1 < 0(R_0 < 1)$.

Proof:(i)if $\lambda_3 < 0$, we can obtain $S^* < \frac{1}{\beta_S b}(\mu_I + \mu_P + bn)(\mu_I + \mu_P + bn - b)$, yield

$$\beta_{S}b[\sqrt{\Delta} + (\Lambda\alpha - \mu_{A}\mu_{S})(\mu_{A} + \mu_{S}) - \alpha\mu_{S}D] + 2\alpha\mu_{S}(\mu_{A} + \mu_{S})(\mu_{I} + \mu_{P} + bn)(b - \mu_{I} - \mu_{P} - bn) < 0$$

we can let $\phi_1 = \sqrt{\Delta} + (\Lambda \alpha - \mu_A \mu_S)(\mu_A + \mu_S) - \alpha \mu_S D < 0$ and $b(n-1) + \mu_I + \mu_P > 0$, then

 λ_3 have negative real part.

(iii) Because $M = \alpha^2 (A^* + S^*)^2 + 2\alpha\mu_S(A^* - S^*) + \mu_S^2 > 0$. We just need $A^* = \frac{D}{\mu_A + \alpha S^*} > S^*$, then M is positive. So, S^* satisfy $\alpha S^2 + \mu_A S - D < 0$, because $\mu_A^2 + 4\alpha D > 0$ and -D < 0, hence, this quadratic equation must have positive root $\frac{-\mu_A + \sqrt{\mu_A^2 + 4\alpha D}}{2\alpha}$. Let $S^* = S_2 < \frac{-\mu_A + \sqrt{\mu_A^2 + 4\alpha D}}{2\alpha}$. We can obtain $\phi_2 < 0$.

A.4 R₀ Using Next Generation Matrix

This section will show in detail how R_0^{NG} was calculated using a Next Generation Matrix. System 2.2 is rewritten as $\mathscr{F} - \mathscr{V}$ where the terms in \mathscr{F} represent the creation of new infectious prion chains. Therefore,

$$\mathscr{F} = \begin{pmatrix} 0\\0\\0\\bZ\\0\\0\\0\\0 \end{pmatrix}, \text{ and } \mathscr{V} = \begin{pmatrix} -\Lambda + \mu_S S + \alpha A S - \mu_A T + \beta_S S P\\\mu_S T - \alpha A S + \mu_A T\\(\mu_P + \mu_I) P + (2n-1)bP\\-\beta_S S P - \beta_R R P + (\mu_P + \mu_I) Z + n(n-1)bP\\\mu_R R + \mu_I \sigma R + \beta_R R P - n(n-1)bP\\\mu_A A + \alpha A S - D \end{pmatrix}.$$

Next, take the jacobian of \mathscr{F} and \mathscr{V} evaluated at the prion-free disease equilibrium $(S^*, T^*, 0, 0, 0, A^*)$:

$$V_{1} = \begin{pmatrix} \alpha A^{*} + \mu_{S} & -\mu_{A} & \beta_{S}S^{*} & 0 & 0 & \alpha S^{*} \\ \alpha - A^{*} & \mu_{A} + \mu_{S} & 0 & 0 & 0 & \alpha - S^{*} \\ 0 & 0 & b(2n-1) + \mu_{I} + \mu_{P} & 0 & 0 & 0 \\ 0 & 0 & b(n-1)n - \beta_{S}S^{*} & \mu_{I} + \mu_{P} & 0 & 0 \\ 0 & 0 & -b(n-1)n & 0 & \mu_{R} + \mu_{I}\sigma & 0 \\ \alpha A^{*} & 0 & 0 & 0 & 0 & \alpha S^{*} + \mu_{A} \end{pmatrix}.$$

Now, F and V can be redefined to reduce the system:

$$F_2 = \begin{pmatrix} 0 & b \\ 0 & 0 \end{pmatrix}, \text{ and}$$
$$V_2 = \begin{pmatrix} b(2n-1) + \mu_I + \mu_P & 0 \\ b(n-1)n - \beta_S S^* & \mu_I + \mu_P \end{pmatrix}.$$

The inverse of V_2 can be written as:

$$V_2^{-1} = \begin{pmatrix} \frac{1}{b(2n-1)+\mu_I + \mu_P} & 0\\ \frac{\beta_S S^* - b(n-1)n}{(\mu_I + \mu_P)(b(2n-1)+\mu_I + \mu_P)} & \frac{1}{\mu_I + \mu_P} \end{pmatrix}$$

Now, find the value $\mathscr{K} = F_2 V_2^{-1}$:

•

$$\mathscr{K} = \begin{pmatrix} \frac{b(\beta_S S^* - b(n-1)n)}{(\mu_I + \mu_P)(b(2n-1) + \mu_I + \mu_P)} & \frac{b}{\mu_I + \mu_P} \\ 0 & 0 \end{pmatrix}.$$

Therefore, $\mathbf{R}_0^{\mathrm{NG}} = \max(\mathrm{Eigenvalues}(\mathscr{K})) =$ $fracb(\beta_S S^* - b(n-1)n)(\mu_I + \mu_P)(b(2n-1) + \mu_I + \mu_P).$

B Derivation of Interferons Formula

Type I IFN-B inhibit prion propagation in infectious cells and animal models

The values of the concentrations of interferon's were taken from experimental data. [12] In the experiments, the concentration is measured in kU/ml and 48h after interferon doses the rate is considered dividing the total amount of prion over the control amount (which does not include the interferon treatment).

Pairs of experimental values:

{concentration(uK/ml), ratio P_I/P_0 } = {0,1}, {0.1, 1.05}, {0.5, 0.93}, {1, 0.945}, {2, 0.90}, {5, 0.65};

B.0.1 Unit Conversion

Since the experimental data is in kU/ml we need to do a conversion of units to obtain the units nM/L.

$$\frac{\mathbf{kU}}{\mathbf{ml}} * \frac{1000\mathbf{U}}{\mathbf{kU}} * \frac{\mathbf{mg}}{2.42 * 10^{7}\mathbf{U}} * \frac{\mathbf{g}}{1000\mathbf{mg}} * \frac{\mu \mathbf{g}}{10^{-6}\mathbf{g}} * \frac{\mu \mathbf{M}}{20027\mu \mathbf{g}} * \frac{\mathbf{10}^{3} \ \mathbf{nM}}{\mu \mathbf{M}} * \frac{1000\mathbf{ml}}{\mathbf{L}}$$
(B.1)

where, 20027g/mol is the molecular mass of an interferon. The value for the conversion was obtained from the data sheet provided by Chemicon (2004) for Mouse Interferon Beta [?].

Then the new list of values are :

Pairs of experimental values {concentration(nM/L), ratio P_I/P_0 }={0.00001,1.00}, {178970,1.05}, {894849,0.93}, {1.7897*10^6, 0.945}, {3.5794*10^6, 0.90}, {8.94849*10^6, 0.65}

B.0.2 Interpolation

The interpolation of the data with a Hill Equation is described by the equation

$$P_{ratio} = \frac{1}{1 + \left(\frac{x}{13989.5 \times 10^3}\right)^{1.44313}} \tag{B.2}$$

Then, assuming that the dosage D and the death rate caused by interferons μ_I is constant,

then we rewrite the total dead rate of the polymer chain $\mu_P + \mu_I$ as $(1 + k)\mu_P$. We assume that the behavior of prions with respect to time (including the treatment) is given by an exponential decay:

$$P(t, I_0) = c_0 e^{-(1+k)\mu_p t}$$
(B.3)

Then, the ratio between the amount of prions when there is treatment and when there is not treatment is given by

$$P_{\text{ratio}}(I) = \frac{c_0 e^{-(1+k)\mu_P t}}{c_0 e^{-\mu_P t}} = \frac{e^{-\mu_P t} e^{-k\mu_P t}}{e^{-\mu_P t}} = e^{-k\mu_P t}$$
(B.4)

Then, if it is assumed a t = 2 days (48 hours) and it is solved for k, it is obtained:

$$k(I) = -\frac{\log(P_{\text{ratio}}(I))}{2\mu_P} \tag{B.5}$$

Where $P_{\text{ratio}}(I)$ is equation 3.9 and it will be a ratio between 0 and 1.

\mathbf{Code}

The code for this calculations was created with Wolfram Mathematica.

(*Values taken from the experiment*)
rateI4={{0.001,1.00},{0.206333,1.05},{1.03167,.93},{2.06333,.945},
{4.12666,.90},{10.3167,.65}};

CvsI= ListPlot[rateI4]

```
hillmodelinter=(1/(1+(x/a)^n));
hillfit=FindFit[rateI4, {hillmodelinter,{n>0,a>0}} ,{a,n}, x]
Show[Plot[hillmodelinter/.hillfit, {x, 0, 9.5*10^6}],
ListPlot[rateI4, PlotRange -> Automatic]]
```

```
(*Assumptions*)
p[t_,co_,k_,uo_]:= co E^(-(1+k)uo t)
po[t_,co_,k_,uo_]:= co E^(-uo t)
inter[Io_,r_,t_]:= Io E^(- r t)
(*k[Io];*)
(*Assume ut and dt ctes?*)
(*Assuming a t=2*)
ratio2= FullSimplify[(p[2,co,k,uo])/(po[2,co,k,uo])]
(*Assuming a t is a variable*)
ratio=FullSimplify[(p[t,co,k,uo])/(po[t,co,k,uo])]
k2=Assuming[
  { { k, uo,f2} \[Element] Reals, f2>0},
  FullSimplify[Flatten[{
    Solve[ratio2==f2,k,Reals]},2]
  ]
]
kt=Assuming[
  { { k, uo} \[Element] Reals, f2>0,f2<1 },
  FullSimplify[
    Solve[ratio2==f2,k,Reals]
  ]
]
kIt=Assuming[
  { { k, uo} \[Element] Reals, f>0,f<1, E^(r t)>0 , Io>0},
  FullSimplify[
```

```
Solve[ratio2==inter[Io,r,t],k,Reals]
]]
```

pratio=hillmodelinter/.hillfit

C Growth Rate

C.1 Getting r

In this section of the appendix we show how to get the growth rate. Asume that S, T, Rand A are initially in a steady state. The system if P and Z is now linear, and will either accumulate or decay exponentially at exponentially at the rate according to the dominant eigenvalue of the Jacobian matrix. After the average size reaches equilibrium, exponential growth in the abundance of PrPSc over time t occurs according to

$$P(t) = P(0)e^{rt}$$
 and $Z(t) = Z(0)e^{rt}$. (C.1)

Taking the equations for P and Z,

$$\frac{dP}{dt} = -(\mu_P + \mu_I)P + bZ - (2n-1)bP,
\frac{dZ}{dt} = \beta SP + \beta RP - (\mu_P + \mu_I)Z - n(n-1)bP,$$
(C.2)

getting the Jacobian that is given by

$$\mathbf{J} = \begin{pmatrix} -(\mu_P + \mu_I) - b(2n-1) & b \\ (S+R)\beta - b(n-1)n & -(\mu_P + \mu_I(t)) \end{pmatrix}$$

The characteristic polynomials of ${\bf J}$ that is given by

$$P(\boldsymbol{\lambda}) = b^2(n-1)n - b(-(2n-1)(\boldsymbol{\lambda} + \mu_I \mu_P) + \beta R + \beta S) + (\boldsymbol{\lambda} + \mu_I + \mu_P)^2$$

and the solutions are given by

$$\boldsymbol{\lambda}_1 = \frac{1}{2} \left(-2bn + \sqrt{b}\sqrt{b + 4\beta R + 4\beta S} + b - 2(\mu_I + \mu_P) \right)$$
$$\boldsymbol{\lambda}_2 = \frac{1}{2} \left(-2bn - \sqrt{b}\sqrt{b + 4\beta R + 4\beta S} + b - 2(\mu_I + \mu_P) \right)$$

taking maximum eigenvalue

$$r = \frac{1}{2} \left(-2bn + \sqrt{b}\sqrt{b + 4\beta R + 4\beta S} + b - 2(\mu_I + \mu_P) \right)$$
$$r = b(1 - n) - \frac{b}{2} + \sqrt{\frac{b^2}{2} + \beta b(S + R)} - (\mu_I + \mu_P)$$
(C.3)

Then, the values of the Prion Disease Equilibria for S^* and R^* are introduced from equations 3.3. Then, the following expression is obtained:

$$r = b(1-n) - \frac{b}{2} + \sqrt{\frac{b^2}{2} + \beta b(S^* + R^*)}$$

$$r = b(1-n) - \frac{b}{2} - (\mu_I + \mu_P) + \sqrt{\frac{b^2}{2} + \beta b \left(\frac{1}{2} \left(\frac{\Lambda}{\mu_S} - \frac{D}{\mu_A + \mu_S} - \frac{\mu_A}{\alpha}\right) + \sqrt{\frac{\Lambda\mu_A}{\alpha\mu_S} + \frac{1}{4} \left(\frac{D}{\mu_A + \mu_S} + \frac{\mu_A}{\alpha} - \frac{\Lambda}{\mu_S}\right)^2}\right)}$$

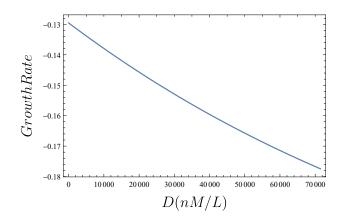


Figure 19: These plots shows the value of the prion growth rate against the concentrations of Pharmacological Chaperon dosages.