Justification of quasi-stationary approximation in models of gene expression of a self-regulating protein

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Abstract

We analyse a model of Hes1 gene transcription and protein synthesis with a negative feedback loop. The effect of multiple binding sites in the Hes1 promoter as well as the dimer formation process are taken into account. We consider three, possibly different, time scales connected with: (i) the process of binding to/dissolving from a binding site, (ii) formation and dissociation of dimers, (iii) production and degradation of Hes1 protein and its mRNA. Assuming that the first two processes are much faster than the third one, using the Tikhonov theorem, we reduce in two steps the full model to the classical Hes1 model. In the intermediate step two different models are derived depending on the relation between the time scales of processes (i) and (ii). The asymptotic behaviour of the solutions of systems are studied. We investigate the stability of the positive steady state and perform some numerical experiments showing differences in dynamics of the considered models.

Keywords: biochemical reaction, Tikhonov theorem, asymptotic analysis, stability, negative feedback loop

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1. Introduction

Regulation of gene expression in eukaryotic cells is one of the most important processes during the life of the cell and the whole organism. How cells work depends on the signals that reach them. The cell’s response is based on changing the expression of genes, and thus on the change in the amount of protein produced. Both silencing and overexpression as defects in gene regulation cause unfavorable changes. Thus, understanding the structure and mechanism of gene expression regulation is necessary to understand the functioning of many biological and chemical processes related mainly to genetic regulation. It is also necessary for understanding the emergence of diseases, and thus effective fight against them. Many cancers arise due to overexpression of major regulatory genes, i.e. genes encoding a regulatory protein.

One such regulatory protein is Hes1 (hairy and enhancer of split 1), which belongs to the helix-loop-helix (bHLH) family of transcription proteins, i.e. DNA-binding proteins in the promoter region or in another region where regulation of transcription processes occurs. Hes1 protein deficiency in mice leads to premature cell differentiation, resulting in defects in brain tissue. In turn, the overexpression of Hes1 has been observed in many cancers, including lung cancer, ovarian cancer and colon cancer as well as germ cell tumors [9]. Hes1 also induces the activation of PARP1 in acute lymphoblastic leukemia, and patient samples during the leukemic crisis showed an elevated level of Hes1. This suggests that Hes1 protein may induce tumor cell growth. In [10] one more possibility of unfavorable Hes1 activity was described. The authors report that Hes1 may promote tumor metastases, including metastases to the tumor bone. This is related to the effect of Hes1 on the proliferation of cancer cells and migration abilities. Moreover, in [10] the relationship between Hes1 and breast cancer was examined in order to identify potential causes of increased invasion and metastasis of breast cancer. The authors examined Hes1 expression using Western blot analyses of freshly isolated breast cancer tissues and observed that patients with low levels of Hes1 expression have an increased survival compared to patients with high levels of Hes1 expression.

It is worth pointing out that many signaling pathways are involved in the regulation of Hes1 gene expression. It is important that Hes1 lies at the crossroads of many signaling pathways. For example Hes1 is regulated by Notch signaling pathway, which is mainly involved in the regulation of hematopoietic cell function and in tumor vasculature.
Targeting in Hes1 can therefore cause fewer side effects as many other target genes of the Notch pathway will remain intact, \[6,9\].

Hes1 protein as a transcriptional repressor, inhibits its own transcription by directly binding to its own promoter, which blocks transcription of Hes1 mRNA (see for instance \[5,13,20\]). When the transcription of Hes1 mRNA is repressed by this negative feedback, Hes1 protein soon disappears because it is rapidly degraded by the ubiquitin-proteasome pathway. Disappearance of Hes1 protein allows then the next round of transcription. In this way, Hes1 protein autonomously starts oscillatory expression induced by a negative feedback loop, see \[5\]. Another example of this is the mechanism of p53 and Mdm2 proteins in which oscillations resulting from stress were observed, \([12]\).

The classical mathematical model which describes the gene expression of Hes 1 protein was proposed by Monk \([11]\). This model includes four basic biochemical processes, i.e. transcription (synthesis), translation (production), protein degradation and its mRNA:

\[
\begin{align*}
\text{change of mRNA concentration} &= \text{transcription rate} - \text{mRNA degradation rate} \\
\text{change of Hes1 protein concentration} &= \text{translation rate} - \text{Hes1 protein degradation rate}
\end{align*}
\]

Moreover, in the model proposed by Monk \([11]\) it was assumed that the intensity of mRNA production is a decreasing function of the concentration of the protein and the transcription time was taken into account. In \([7]\) this suppression function was assumed to be a Hill function with the Hill coefficient greater than two (due to the assumption that dimer binding to DNA is co-operative one). There are several approaches in the literature for modeling the gene expression of the hes1 protein. Hirata in \([5]\) postulated the existence of a third non-linear component in the Hes1 model that causes oscillations. However, most models describing small autoregulation networks, such as the Hes1 model, are based on delay differential equation (DDE) and the oscillatory behaviour is caused by the delay in transcription and/or translation processes due to the Hopf bifurcation. In Bernard et al. \([2]\) a version of the Monk Hes1 model with delay in transcription process was considered, while in \([3]\) we have studied the model with delay in both transcription and translation processes. In particular, we showed that the crucial factor for the appearance of oscillations is a sum of time delays in transcription and translation. Moreover, the direction of appearing Hopf bifurcation was calculated. In addition to the models based on delay differential equations, there exist models of the Hes1 regulatory pathway focusing on the spatial aspect, i.e. transport between the nucleus and the cytoplasm \([15,16,17,19]\). Some mathematical analysis of this model can be found in \([4,8]\).

In this paper we consider a modification of a classical Hes1 gene expression model. Our main goal is to justify mathematically the form of the classical Hes1 model and to show that the stability of the system depends on the number of binding sites that is places at the Hes1 promoter, at which the complexes of Hes1 protein bind blocking the protein transcription.

It is worth pointing out that before a protein binds to its own DNA, it creates a dimer that inhibits the transcription of its mRNA. For this reason, in the system we propose, we take into account the reaction describing the binding of the dimer to the binding site in the regulatory region of the gene. Since DNA promoter usually has multiple binding sites, \([20]\), their number has a strong effect on the dynamic behaviour of the Hes1 protein system. The case of three binding sites was studied earlier by Zeiser \([22]\). The full model that describes dimer formation as well as many binding sites in the protein promoter leads to the system of \(n + 3\) ordinary differential equations (ODEs), where \(n\) is the number of the binding sites. Due to the complexity of this process, we will simplify it using quasi-stationary approximation. Namely, we assume that the process of dimers binding to and dissociating from DNA promoter is much faster than other considered processes, and thus we consider that the probability that DNA is in active state can be described as a function of proteins’ dimers. Mathematically, this means that we can write a small parameter (or parameters) on the left-hand side of equations, and then use the Tikhonov theorem, see \([21]\). The Tikhonov method reduces variables of the complex systems using the assumption that some parameters are small. We emphasize once again that there are two processes here: dimer formation and binding to the promoter, and that both may have different time scales. In this way, simplification of the entire system to the classical one, we can follow two (or even three) different ways, depending on the time scales of the processes mentioned earlier, see Fig. 1.

The paper is organized as follows. In Section 2 we derive a general gene expression model, we formulate its basic mathematical properties and we prove the existence of a unique positive steady state. Next, using quasi-stationary approximation we formally obtain three simplified models under different assumptions on time scales. In Section 3 using Tikhonov theorem we rigorously justify the form of reduced models formally derived in Section 2. In Section 4 we study stability of the positive steady state of the reduced systems. The paper is concluded by numerical simulations and discussions presented in Section 5.
2. Gene expression models

2.1. Model derivation

Hes1 has at least three binding sites (see for instance [20]). If the Hes1 dimer binds to one of these sites it blocks transcription. Let us derive equations that would describe probabilities that a given number of sites are occupied. To this end, we assume that the concentration of Hes1 dimers is given by $y_2$. We construct equations for the change of the concentration of Hes1 dimers later. We show that it is enough to consider probabilities that $j$ sites are occupied and the particular configuration of free-occupied sites is not important as long as we assume that probability that Hes1 dimer binds to a free site does not depend on a particular site and is the same for all $n-j$ free sites (but the probability may depend on the number of free/occupied sites).

To finish argumentation we need some notation to be introduced. Let $\Sigma$ denote the set of all permutations of the $n$-element set. Denote by $e_j^n$ the vector $(1, \ldots, 1, 0, \ldots, 0)$, where 1 is on the first $j$th positions and 0 on the last $(n-j)$th positions. Let

$$e_{jk}^n(k) = 1 - e_j^n(k), \quad e_{jk}^n(\ell) = e_j^n(\ell), \quad \ell \in \{1, 2, \ldots, n\} \setminus \{k\}.$$ 

We see that $e_{jk}^n$ denote the vector that differs from $e_j^n$ only on the $k$th positon. For $\sigma \in \Sigma$ let $x_{\sigma(e_j^n)}$ denote the probability that the configuration of occupied and free sites is given by a vector $\sigma(e_j^n)$, where 1 on the $\ell$th coordinate of the vector $\sigma(e_j^n)$ means that the $\ell$th site is occupied and 0 means that it is free. Let us also denote by $k_j y_2$ the probability that a Hes1 dimer binds to one of free sites given that $j$ is occupied and let $\gamma_j$ be an intensity of dissolving of a Hes1 dimer assuming that there are $j$ occupied sites. The change of the probability $x_{\sigma(e_j^n)}$ (we assume that $1 \leq j \leq n-1$ for simplicity), is due to one of the following actions (see Fig. 2):

- the Hes1 dimer may bind to one of $n-j$ free sites, to each with probability $k_j y_2/(n-j)$;
- one of $j$ bounded dimers may dissolve, each with intensity $\gamma_j$.

![Figure 2: The scheme of the process of binding dimers to the promoter and dissociating them. The crosses indicate occupied sites.](image)
Using this assumptions and the mass action law we write
\[ x'_{\sigma(e^j)} = \frac{j}{n-1} n + j - 1 y_2 x_{\sigma(e^j)} + \sum_{j=1}^{n} y_{j+1} x_{\sigma(e^j)} - (k_j y_2 + j y_j) x_{\sigma(e^j)}, \] (2.1)
for \( 1 \leq j \leq n - 1 \). For the cases \( j = 0 \) and \( j = n \), we have
\[ x'_{\sigma(e^0)} = \sum_{j=1}^{n} y_1 x_{\sigma(e^j)} - k_0 y_2 x_{\sigma(e^0)}, \]
\[ x'_{\sigma(e^n)} = \sum_{j=1}^{n} k_{n-1} y_2 x_{\sigma(e^j)} - n y_n x_{\sigma(e^n)}. \] (2.2)
Now, let us denote
\[ x_j = \sum_{\sigma(e^j)} x_{\sigma(e^j)}. \]
We note that \( \sigma(e^0) \) is one of \( \sigma(e^0, k, k') \) if \( 1 \leq k \leq j \) and one of \( \sigma(e^0, k, k') \) if \( j + 1 \leq k \leq n \). This, together with the fact that Eqs. (2.1) and (2.2) are linear with respect to \( \sigma(e^0) \), leads to the following equations
\[ x'_0 = y_1 x_1 - k_0 y_2 x_0, \]
\[ x'_j = k_{j-1} x_{j-1} y_2 + (j + 1) y_{j+1} x_{j+1} - (k_j y_2 + j y_j) x_j, \]
\[ 1 \leq j \leq n - 1, \]
\[ x'_n = k_{n-1} y_2 x_n - n y_n x_n \]
and
\[ y'_1 = 2 y_1 y_2 - 2 k_1 y_1^2 + r_z - \delta y_1, \]
\[ y'_2 = - \sum_{j=1}^{n-1} k_j x_j y_2 + \sum_{j=1}^{n} j y_j x_j - \gamma y_2 + k_j y_1^2, \]
\[ z' = r_z x_0 - \delta y, \]
where \( \gamma \) denotes an intensity of a formation of Hes1 dimers, \( k_j \) denotes an intensity of a formation of Hes1 dimers, \( r_z \) and \( r_y \) production rates of Hes1 and its mRNA, respectively, while \( \delta y \) and \( \delta z \) are degradation rates of Hes1 and its mRNA, respectively.

### 2.2. Non-dimensionalisation and basic mathematical properties

Before analysing model (2.3)-(2.4) we express it in non-dimensional terms, thereby reducing the number of parameters.

The right-hand side of system (2.3)-(2.4) is a polynomial, thus existence of a unique solution to this system is immediate. The fact that \( x_0 + x_1 + \cdots + x_n = 1 \) implies that system (2.3)-(2.4) can be reduced to a system of \( n + 3 \) equations. However, we find it convenient to write the system in the present perturbed form. We show that this problem can be reduced to a lower-dimensional problem. We choose the scaling in such a way, that the positive steady state (which is a unique steady state of system (2.3)-(2.4), as we prove later) has a very simple coordinates. In order to do that we proceed in the following manner. Let \( q \) be the positive solution to the following equation
\[ \frac{\delta y \delta z}{r_y r_z} q = \frac{1}{1 + \sum_{j=1}^{n} \frac{k_j}{\gamma_j} \left( \frac{k_j}{\gamma_j} q^2 \right)}. \]
We introduce non-dimensional quantities putting
\[ \bar{x}_j = x_j, \quad \bar{y}_1 = \frac{y_1}{q}, \quad \bar{y}_2 = \frac{y_2}{k_0 q^2}, \quad \bar{z} = \frac{r_z}{\delta y} q, \quad \tau = k_0 q^2 t \]
and
\[ k = \frac{2}{q}, \quad \delta y = \frac{\delta y}{k_0 q^2}, \quad \delta z = \frac{\delta z}{k_0 q^2}, \quad \bar{y}_j = \gamma_j \frac{\gamma y}{k_0 k_j q^2}, \quad \bar{e}_1 = \frac{\gamma y}{k_0}, \]
\[ \bar{e}_2 = \frac{k_j}{k_0}, \quad r_0 = \frac{r_y r_z}{\delta y \delta z q} \]
\[ \theta = \frac{k_0}{\gamma y}, \quad \bar{e}_2 = \frac{k_j q^2}{\gamma y}. \]
for $0 \leq j \leq n$. Note that $\hat{k}_0 = 1$. Nevertheless, we keep writing $\hat{k}_0$ to get natural and (in some sense) more symmetric formulas.

**Remark 1.** We observe, that due to the definitions of $q$ and $r_0$ dimensionless parameters fulfil the following equality

$$r_0 = 1 + \sum_{j=1}^{n} \frac{1}{\hat{k}_j} \hat{k}_0 \ldots \hat{k}_{j-1}.$$  

(2.5)

For notational simplicity we drop the tilde on $x_j, y_1, y_2, z, k_j$ and $y_j (0 \leq j \leq n)$ and, in consequence, the non-dimensional version of the system (2.3)–(2.4) reads as follows

$$\varepsilon_k \hat{x}_0 = \gamma_1 x_1 - k_0 x_0 y_2,$$

$$\varepsilon_k x_j' = k_{j-1} x_{j-1} y_2 + (j + 1) y_{j+1} x_{j+1} - (k_j y_2 + j y_j) x_j, \quad 1 \leq j \leq n - 2,$$

$$\varepsilon_k x_{n-1}' = k_{n-2} x_{n-2} y_2 + n y_n \left(1 - \sum_{j=0}^{n-1} x_j\right) - (k_{n-1} y_2 + (n-1) y_{n-1}) x_{n-1},$$

$$y_j' = k y_2 - y_j^2 + \delta_1 (z - y_j),$$

$$\varepsilon_k y_n' = \theta \left(- \sum_{j=0}^{n-1} k_j x_j y_2 + \sum_{j=1}^{n-1} j y_j x_j + n y_n \left(1 - \sum_{j=0}^{n-1} x_j\right)\right) - y_2 + y_j^2,$$

$$\gamma' = \delta_2 (r_0 z - \gamma).$$

(2.6)

From now on we deal with dimensionless model.

### 2.2.1. Uniqueness, existence, and non-negativity of solutions and existence of a unique positive steady state

**Theorem 2.** The solutions to (2.6) exist are nonnegative, unique and defined for all $t$. Every set

$$\Omega = \{(x_0, \ldots, x_n, y_1, y_2, z) \in \mathbb{R}^{n+3} : 0 \leq x_j, \sum_{j=0}^{n-1} x_j \leq 1, 0 \leq y_1 \leq \bar{y}_1, 0 \leq y_2 \leq \bar{y}_2, 0 \leq z \leq r_0, 0 \leq j \leq n - 1\}$$

such that constants $\bar{y}_1, \bar{y}_2$ fulfil

$$\bar{y}_1 = r_0 + \frac{k}{\delta_1} \theta, \quad \bar{y}_2 = \bar{y}_1^2 + \theta_1, \quad \theta_1 = \theta \sum_{j=1}^{n} j y_j,$$

(2.7)

is invariant for the evolution system (2.6).

**Proof.** Since $\sum_{j=0}^{n} x_j(t) = 1$ and all variables are positive we immediately have $0 \leq x_0(t) \leq 1$. Thus, from the last equation of (2.6), we get

$$\gamma' (t) \leq \delta_2 (r_0 - \gamma) \Rightarrow \gamma (t) \leq \max\{\gamma (0), r_0\}.$$

If the initial condition is from $\Omega$ then $\gamma (t) \leq r_0$. For the similar reason, the equations for $y_1$ and $y_2$ can be estimated as

$$y_1' \leq k (y_2 - y_1^2) + \delta_1 (r_0 - y_1), \quad y_2' \leq \theta \left(\sum_{j=1}^{n} j y_j x_j + n y_n \left(1 - \sum_{j=0}^{n-1} x_j\right)\right) - y_2 + y_1^2 \leq \theta - y_2 + y_1^2,$$

where we used the definition of $\theta_1$. Let draw two curves

$$y_2 = y_1^2 + \theta_1, \quad y_2 = y_1^2 + \frac{\delta_1}{k} (y_1 - r_0),$$

(2.8)

in the $(y_1, y_2)$ plane (see the solid red and dashed blue curve, respectively, in Fig. 3). Note, that $y_1$ decreases below the second curve while $y_2$ decreases above the first one. Note also that these two curves intersect at the point

$$\bar{y}_1 = r_0 + \frac{k}{\delta_1} \theta_1 > 0.$$

The line connecting the point $(0, \bar{y}_2)$ with $(\bar{y}_1, \bar{y}_2)$ is above the curve $y_2 = y_1^2 + \theta_1$ (the solid red line in Fig. 3) and thus, $y_2' < 0$. The line connecting the point $(\bar{y}_1, 0)$ with $(\bar{y}_1, \bar{y}_2)$ is below the curve $y_2 = y_1^2 + \frac{\delta_1}{k} (y_1 - r_0)$ (the dashed blue line in Fig. 3) and thus, $y_2' > 0$. This shows that $y_1(t), y_2(t)$ cannot escape the region bounded by these hyperplanes, which completes the proof. 

\[ \square \]
Combing this we have the differential equations that describe concentrations of Hes1 mRNA and Hes1 proteins. In this section we reduce model (2.6) to the classical differential equations and obtain a unique steady state.

**Remark 3.** For any constants \( \hat{\gamma}_1, \hat{\gamma}_2 \) fulfills \( \hat{\gamma}_1 < \check{\gamma}_1, \hat{\gamma}_1^2 + \theta \gamma < \check{\gamma}_2 < \hat{\gamma}_1^2 + \frac{\theta}{\phi} (\hat{\gamma}_1 - r_0) \) a set

\[
\tilde{\Omega} = \left\{ (x_0, \ldots, x_{n-1}, y_1, y_2, z) \in \mathbb{R}^{n+3} : 0 \leq x_i, \sum_{j=0}^{n-1} x_j \leq 1, 0 \leq y_1 \leq \hat{\gamma}_1, 0 \leq y_2 \leq \check{\gamma}_2, 0 \leq z \leq r_0, 0 \leq i \leq n - 1 \right\}
\]

is invariant under the evolution of system (2.6). Note also that the inequality \( \hat{\gamma}_1 < \check{\gamma}_1 \) implies that \( k \theta \gamma < \delta (\hat{\gamma}_1 - r_0) \).

**Proposition 4.** There exists exactly one non-negative steady state \( (\hat{x}_0, \hat{x}_1, \ldots, \hat{x}_{n-1}, \hat{y}_1, \hat{y}_2, \bar{z}) \) of system (2.6), where

\[
\hat{x}_0 = \frac{1}{r_0}, \quad \hat{x}_j = \frac{1}{j!} \frac{k_0 k_1 \cdots k_{j-1}}{\gamma_j y_2 \cdots y_j} \left[ \frac{1}{r_0} - 1 \right] \quad 1 \leq j \leq n - 1, \quad \hat{y}_1 = 1, \quad \hat{y}_2 = 1, \quad \bar{z} = 1.
\]

**Proof.** Note that the first \( n \) equations are linear with respect to \( x_0, x_1, \ldots, x_{n-1} \) if \( y_2 \) is fixed. Thus, looking for a steady state \( (\hat{x}_0, \hat{x}_1, \ldots, \hat{x}_{n-1}, \hat{y}_1, \hat{y}_2, \bar{z}) \) of system (2.6) we easily get

\[
\hat{x}_j = \frac{1}{j!} \frac{k_{j-1}}{\gamma_j} \hat{y}_2 \hat{x}_{j-1} = \frac{1}{j} \frac{k_0 k_1 \cdots k_{j-1}}{\gamma_j y_2 \cdots y_j} \hat{y}_2 \hat{x}_0, \quad j = 1, 2, \ldots, n,
\]

and \( \hat{x}_n = 1 - (\hat{x}_0 + \hat{x}_1 + \cdots + \hat{x}_{n-1}) \).

Now, it is easy to see that the following identity

\[
\hat{x}_0 = \frac{1}{1 + \sum_{j=1}^{n} \frac{1}{j!} \frac{k_0 k_1 \cdots k_{j-1}}{\gamma_j y_2 \hat{y}_2}}
\]

holds. From the last equation of (2.6) we deduce that

\[
\bar{z} = r_0 \hat{x}_0.
\]

The equation for \( y_2 \) implies that \( \hat{y}_2 = \hat{y}_1^2 \) (the expression that is multiplied by \( \theta \) is equal to zero due to (2.11)) and therefore

\[
\hat{y}_1 = \bar{z}.
\]

Combing this we have

\[
\hat{y}_1 = \frac{r_0}{1 + \sum_{j=1}^{n} \frac{1}{j!} \frac{k_0 k_1 \cdots k_{j-1}}{\gamma_j y_2 \hat{y}_2}}.
\]

We see at once that equation (2.12) has a unique solution since its right-hand side is a positive decreasing function of \( y_1 \), its left-hand side is a linear increasing function taking value 0 at \( \hat{y}_1 = 0 \). Due to (2.5) it is immediate that \( \hat{y}_1 = 1 \) solves (2.12) and then formulas (2.9) follow.

**2.3. Formal derivation of reduced model**

The classical Hes1 gene expression model proposed by Monk [11] is a system of only two differential equations that describes concentrations of Hes1 mRNA and Hes1 proteins. In this section we reduce model (2.6) to the classical Hes1 model in the two consecutive steps by means of the quasi-stationary approximation. We consider that the complex
To this end, we set $\varepsilon$ to obtain the system $\phi$. Treating $\varepsilon$, we see at once that $x$ depends on whether $\varepsilon_2 \ll \varepsilon_1$ or $\varepsilon_1 \ll \varepsilon_2$ or $\varepsilon_1 \approx \varepsilon_2$. The last case is more complex and will be considered elsewhere.

In the first case we set $\varepsilon_2 \approx 0$ and $\varepsilon_1 > 0$ obtaining the system in which the concentration of dimers is always in a stationary level. Then we put $\varepsilon_1 \approx 0$ which reduces the system to the classical one. In the second case we first set $\varepsilon_1 \approx 0$ and $\varepsilon_2 > 0$ obtaining system in which the probability that DNA is not blocked as a function of hes1 dimers concentration. Then we set $\varepsilon_2 \approx 0$ obtaining again the classical system. These two approaches are schematically illustrated in Fig.1.

2.3.1. Derivation of a simplified model under the assumption that dimer dynamics is much faster than other ones.

We start from the reduction of the model (2.6) assuming that dynamics of dimer formation and dissociation is much faster than other processes. In this case $\varepsilon_2$ is very small, thus we consider $\varepsilon_2 \rightarrow 0$. Now, we derive a simplified model. To this end, we set $\varepsilon_2 \approx 0$ and we calculate $y_2$ in dependence on other variables obtaining

$$y_2 = \frac{y_1^2 + \theta \sum_{j=1}^{n-1} jy_j x_j + \theta n y_1 (1 - \sum_{j=0}^{n-1} x_j)}{1 + \theta \sum_{j=0}^{n-1} k_j x_j} = \varphi(x, y_1),$$

(2.13)

where $x = (x_0, \ldots, x_{n-1})$. The expression above describes the stationary concentration of Hes1 dimers when other quantities are given. The rest of the equations of (2.6) are exactly as before, i.e.

$$\varepsilon_1 x_0' = \gamma_1 x_1 - k_0 x_0 \varphi(x, y_1),$$

$$\varepsilon_1 x_j' = k_{j-1} x_{j-1} \varphi(x, y_1) + (j + 1) y_j x_{j+1} - (k_j x_j + j y_j) x_j, \quad 1 \le j \le n - 2,$$

$$\varepsilon_1 x_{n-1}' = k_{n-2} x_{n-2} \varphi(x, y_1) + n y_n \left(1 - \sum_{j=0}^{n-1} x_j\right) - (k_{n-1} x_1 + (n - 1) y_{n-1}) x_{n-1},$$

(2.14)

and

$$y_j' = k (\varphi(x, y_1) - y_1^2) + \delta_1 (z - y_1),$$

$$z' = \delta_2 (\nu_1 x_0 - z),$$

where $\varphi(x, y_1)$ is given by (2.13).

2.3.2. Derivation of a simplified model under the assumption that dynamics of free and occupied sites is much faster than other ones.

In order to simplify the system (2.6), the first $n$-equations can be reduced to algebraic equations. Setting $\varepsilon_1 \approx 0$ we obtain the system

$$\gamma_1 x_1 - k_0^0 x_0 = 0,$$

$$k_{j-1} x_{j-1} + (j + 1) y_j x_{j+1} - (k_j x_j + j y_j) x_j = 0, \quad 1 \le j \le n - 2,$$

$$k_{n-2} x_{n-2} + n y_n \left(1 - \sum_{j=0}^{n-1} x_j\right) - (k_{n-1} x_1 + (n - 1) y_{n-1}) x_{n-1} = 0.$$  (2.15)

Treating $y_2$ as a parameter we solve (2.15) obtaining

$$x_0 = \psi(y_2),$$

$$x_j = \frac{k_0 \ldots k_{j-1}}{j!} y_j^2 \psi(y_2), \quad 1 \le j \le n - 1,$$

(2.16)

where

$$\psi(y_2) = \frac{1}{1 + \sum_{j=1}^{n} \frac{k_j}{\gamma_j \gamma_j} y_j^2}.$$

(2.17)

Moreover, we see at once that

$$- \sum_{j=0}^{n-1} k_j x_j y_2 + \sum_{j=1}^{n-1} j y_j x_j + n y_n \left(1 - \sum_{j=0}^{n-1} x_j\right) = 0,$$

(2.18)
3 Justification of quasi-stationary approximation using the Tikhonov theorem

3.1. The Tikhonov theorem

Assume that in a system of ordinary differential equations there exists a subsystem, which dynamics is much faster than others equations. If such “fast” subsystem can be distinguished, then the naive thinking is that this fast subsystem is close to its stationary state so we can eliminate from one to several variables from the full system and replace it by some algebraic equations. However, it is clear that such approach need not always to be true — imagine for example that the steady state of the fast subsystem is unstable. In this Section, we remind a mathematical theory that justify the approximation described above. It is based on the Tikhonov theorem of 1952, (see [1, 21]), which states that as small parameter $\varepsilon > 0$ converges to zero, the solution of the full system approaches the solution of the degenerate (slow) system. More precisely, the Tikhonov theorem implies that the solutions of full system can be approximated by the solutions of the reduced system. In order to make the paper clearer we cite here the Tikhonov theorem together with needed assumptions following [1]. We consider the following system of ODEs with one small parameter

$$
\begin{align*}
    u'(t) &= F(u, v), \quad u(0) = u_0, \\
    \varepsilon v'(t) &= G(u, v), \quad v(0) = v_0.
\end{align*}
$$

(3.1)

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Finally, we get the following reduced system

$$
\begin{align*}
    y_1' &= k(y_2 - y_1^2) + \delta_1(z - y_1), \\
    \varepsilon_2 y_2' &= y_1^2 - y_2, \\
    z' &= \delta_2(r_0\psi(y_2) - z), \\
    y_2 &= y_1^2,
\end{align*}
$$

(2.18)

where the function $\psi$ is given by (2.17).

2.3.3. Reduction of system (2.14) and (2.18) into the classical Hes1 gene expression model

Here, we derive the classical Hes1 gene expression model starting from (2.14) and (2.18). We first consider (2.18).

We observe that the left-hand side of the second equation of (2.18) is multiplied by a small parameter $\varepsilon_2$, since the creation and dissociation of Hes1 dimers are much faster than translation and transcription processes. We derive now, the dependence of $y_2$ on $y_1$. So, putting $\varepsilon_2 \approx 0$ and using the second equation of (2.18), we get

$$
y_2 = y_1^2,
$$

which leads to the reduced system

$$
\begin{align*}
    y_1' &= \delta_1(z - y_1), \\
    z' &= \delta_2(r_0\psi(y_1^2) - z).
\end{align*}
$$

(2.19)

On the other hand, putting $\varepsilon_1 \approx 0$ in (2.14) and proceeding as in Section 2.3.2 we also arrive to system (2.19). Model (2.19) is well known in the literature and we only cite the stability result that can be found, for example, in [3]. We cite it here for clarity.

**Proposition 5.** If the function $\psi$ is strictly decreasing, the positive steady state of the system (2.19) is locally asymptotically stable.

Of course, the function $\psi$ given by (2.17) is a decreasing function.

**Remark 6.** In the classical Hes1 model (see [7]), the term $\psi(y_1^2)$ is replaced by the Hill function

$$
\psi_h(y_1^2) \approx \frac{a^h}{a^h + y_1^2},
$$

(2.20)

where $h$ is the Hill coefficient.

3. Justification of quasi-stationary approximation using the Tikhonov theorem

Assume that in a system of ordinary differential equations there exists a subsystem, which dynamics is much faster than others equations. If such “fast” subsystem can be distinguished, then the naive thinking is that this fast subsystem is close to its stationary state so we can eliminate from one to several variables from the full system and replace it by some algebraic equations. However, it is clear that such approach need not always to be true — imagine for example that the steady state of the fast subsystem is unstable. In this Section, we remind a mathematical theory that justify the approximation described above. It is based on the Tikhonov theorem of 1952, (see [1, 21]), which states that as small parameter $\varepsilon > 0$ converges to zero, the solution of the full system approaches the solution of the degenerate (slow) system. More precisely, the Tikhonov theorem implies that the solutions of full system can be approximated by the solutions of the reduced system. In order to make the paper clearer we cite here the Tikhonov theorem together with needed assumptions following [1]. We consider the following system of ODEs with one small parameter

$$
\begin{align*}
    u'(t) &= F(u, v), \quad u(0) = u_0, \\
    \varepsilon v'(t) &= G(u, v), \quad v(0) = v_0.
\end{align*}
$$

(3.1)

System (3.1) consist of two subsystems: equations for $u$ (which is usually called a slow system, and equations for $v$, which is called fast subsystem. To formulate the Tikhonov theorem we need the following assumptions:
(A1) $F: \Omega \to \mathbb{R}^n$ and $G: \Omega \to \mathbb{R}^m$ are continuous and satisfy the Lipschitz condition in $\Omega$, where $\Omega = \overline{U} \times V$ is a subset of $\mathbb{R}^{n+m}$, where $\overline{U}$ is a compact set in $\mathbb{R}^n$ and $V$ is a bounded open set in $\mathbb{R}^m$.

(A2) for any $u \in \overline{U}$ there exists an isolated solution $v = \phi(u) \in V$ of the algebraic equation $G(u,v) = 0$ and $\phi$ is continuous.

(A3) for any $u \in \overline{U}$ treated as a parameter the solution of the initial layer equation $v'(t) = G(u,v)$ is asymptotically stable (uniformly with respect to $u$),

(A4) the function $u \mapsto F(u,\phi(u))$ satisfies the Lipschitz condition with respect to $u$ in $\overline{U}$ and there exists a unique solution $\bar{u}(t)$ of the reduced system

$$u'(t) = F(u,\phi(u)), \quad u(0) = u_0$$

such that $\bar{u}(t) \in \text{Int} \overline{U}$ for all $t \in (0,T)$,

(A5) $v_0$ belongs to the region of attraction of the point $\phi(u_0)$, where $G(u_0,\phi(u_0)) = 0$, i.e. the solution $\hat{v} = \hat{v}(t)$ of the initial problem

$$v'(t) = G(u_0,v), \quad v(0) = v_0$$

satisfies $\lim_{t \to \infty} \hat{v}(t) = \phi(u_0)$.

Theorem 7 (Tikhonov). Let $T > 0$ be an arbitrary number. Under the assumptions (A1)–(A5) there exists $\varepsilon_0 > 0$, such that for any $\varepsilon \in (0,\varepsilon_0]$ there exists a unique solution $(u_\varepsilon(t), v_\varepsilon(t))$ of the full system (3.1) on $[0,T]$ and

$$\lim_{\varepsilon \to 0} u_\varepsilon(t) = \bar{u}(t), \quad t \in [0,T],$$

$$\lim_{\varepsilon \to 0} v_\varepsilon(t) = \hat{v}(t), \quad t \in (0,T),$$

where $\bar{u}(t)$ is the solution of the reduced problem (3.2) and $\hat{v}(t) = \phi(\bar{u}(t))$.

The Tikhonov theorem gives the conditions under which the solution $(u_\varepsilon(t), v_\varepsilon(t))$ of system (3.1) converges to $(\bar{u}(t), \hat{v}(t))$, where

- $\hat{v}$ is the solution of the algebraic equation $0 = G(u,v)$,
- $\bar{u}$ is the solution to (3.2) obtained from the first equation of system (3.1) by substituting a known quasi-stationary solution $\hat{v}$ instead of $v$.

Now, using the Tikhonov theorem we prove that for $\varepsilon_1$ small enough solution to system (2.18) (the model with dimers) approximates solution to system (2.6) (the full model), and solution to (2.19) (the classical Hes1 model) approximates solutions to (2.14) (the model without dimers), see Fig. 1. We also show that for $\varepsilon_2$ small enough solution to (2.14) (the model without dimers) approximates solution to system (2.6) (the full model), and solution to (2.19) (the classical Hes1 model) approximates solutions to (2.18) (the model with dimers), see the upper left and lower right arrows at Fig. 1.

To this end we need to introduce some notation. Let

$$\Omega_x = \{(x_0, \ldots, x_{n-1}) \in \mathbb{R}^n : 0 \leq x_j \leq 1, \sum_{j=0}^{n-1} x_j \leq m \leq n - 1\},$$

$$\Omega_y = \{y_2 \in \mathbb{R} : y_2 \leq \tilde{y}_2\},$$

$$\Omega_w = \{(y_1, z) \in \mathbb{R}^2 : 0 \leq y_1 \leq \tilde{y}_1, 0 \leq z \leq r_0\},$$

where $\tilde{y}_1$ and $\tilde{y}_2$ are given by formula (2.7). We rewrite the full system (2.6) in the following way

$$\varepsilon_1 x' = f(x,y_2),$$

$$\varepsilon_2 y_2' = g(x,y_2,w),$$

$$w' = h(x,y_2,w),$$

with initial condition

$$x(0) = \hat{x}, \quad y_2(0) = \hat{y}_2, \quad w(0) = \hat{w},$$

3 JUSTIFICATION OF QUASI-STATIONARY APPROXIMATION USING THE TIKHONOV THEOREM
where \( x = (x_0, x_1, \ldots, x_{n-1}) \), \( w = (y_1, z) \), \( \ddot{w} = (\dot{y}_1, \dot{z}) \) and the functions \( f, g, h \) are given by the following forms:

\[
\begin{align*}
    f(x, y_2) &= \begin{bmatrix} f_0(x, y_2) \\ \vdots \\ f_{n-1}(x, y_2) \end{bmatrix} = \begin{bmatrix} y_1 x_1 - k_0 x_0 y_2 \\ k_{j-1} x_{j-1} y_2 + (j + 1) \gamma_{j+1} x_j + 1 - (k_{j} y_2 + f_j) x_j \\ \vdots \\ k_{n-2} x_{n-2} y_2 + n \gamma_n (1 - \sum_{j=0}^{n-1} x_j) - (k_{n-1} y_2 + (n-1) \gamma_{n-1}) x_{n-1} \end{bmatrix}, \\
g(x, y_2, w) &= \theta \begin{bmatrix} -\sum_{j=0}^{n-1} k_j x_j + 2 \sum_{j=1}^{n-1} f_j x_j + n \gamma_n (1 - \sum_{j=0}^{n-1} x_j) \end{bmatrix} - y_2 + y_1^2 \\
h(x, y_2, w) &= \begin{bmatrix} k(y_2 - \bar{y}_1^2) + \delta_1(z - y_1) \\ \delta_2(n x_0 - z) \end{bmatrix}.
\end{align*}
\]

Remark 8. We observe that the functions \( f : \Omega_{\epsilon} \times \Omega_{\epsilon} \to \mathbb{R} \) and \( g, h : \Omega_{\epsilon} \times \Omega_{\epsilon} \times \Omega_{\epsilon} \to \mathbb{R} \) given by (3.5)–(3.7) are smooth.

Remark 9. Recall that the set \( \Omega = \Omega_{\epsilon} \times \Omega_{\epsilon} \times \Omega_{\epsilon} \) is invariant (see Theorem 2). In fact, it is even a trapping region, i.e. the solutions are contained in \( \text{Int} \Omega \) for \( t > 0 \), because the respective inequalities in the proof of the invariance of \( \Omega \) become strict for \( t > 0 \) (the vector field points inward everywhere on the boundary of \( \Omega \)). For example, if \( x_0 = 0 \) we have \( x_0' = \frac{2}{a_1} > 0 \) and for \( x_0 = 1 \) we deduce \( x_0' = -\frac{2}{a_1} < 0 \). It means that the vector field is pointing to the left, so trajectories cannot leave the interior of the domain. Similar arguments apply to the variables \( y_1 \) and \( z \). Namely, for \( y_1 = 0, y_1' = k y_2 + \delta_1 z > 0 \) for \( |y_1| = 0 \) but for \( y_1 = 1, \bar{y}_1(\dot{y}_1) = 0 \), where \( \bar{y}_1 = r_0 + \frac{1}{s_0} \theta, \) since \( \bar{y}_1 > 1 \), the following inequality is satisfied:

\[
\begin{align*}
    y_1'(\bar{y}_1) &= \frac{k \theta}{1 + \theta x_0} \left( \gamma_1 (1 - x_0) - x_0 \bar{y}_1^2 \right) + \delta_1 (z - \bar{y}_1) \\
    &< \frac{k \theta}{1 + \theta x_0} \left( \gamma_1 (1 - x_0) - x_0 \bar{y}_1 \right) + \delta_1 (z - \bar{y}_1).
\end{align*}
\]

Then, by \( \varepsilon \leq r_0, y_1'(\bar{y}_1) < 0 \). Moreover, for \( z = 0 \) we have \( \zeta' = \delta_2 r_0 x_0 > 0 \), while \( z = r_0 \) we get \( \zeta' = \delta_2 r_0 (x_0 - 1) < 0 \). Combining these we deduce that the solution \( (x_0, \ddot{w}) \) belongs to \( \text{Int} \Omega \).

Now we prove two results on the global stability of the reduced form of (3.3) (i.e. with or without dimers). First, we consider the model which describes changes of concentration of Hes1 dimers.

Proposition 10. Let \( g \) be the function defined by formula (3.6). For any fixed \( x \in \Omega_{\epsilon} \) and \( w \in \Omega_{\epsilon} \), there exists exactly one non-negative steady state of equation \( \varepsilon_2 y_2' = g(x, y_2, w) \) which is globally asymptotically stable in \( \Omega_{\epsilon} \) (uniformly with respect to \( (x, w) = (x_0, \ldots, x_{n-1}, y_1, z) \)).

Proof. Note that we can rewrite the function \( g \) in the following way

\[
g(x, y_2, w) = g(x_0, x_1, \ldots, x_{n-1}, y_2, y_1, z) = -\left( 1 + \theta \sum_{j=0}^{n-1} k_j x_j \right) y_2 + \theta \sum_{j=1}^{n-1} f_j x_j + n \gamma_n \left( 1 - \sum_{j=0}^{n-1} x_j \right) + y_1^2.
\]

Clearly, the equation \( \varepsilon_2 y_2' = g(x, y_2, w) \) is linear with respect to \( y_2 \), as we consider fixed \( x \) and \( w = (y_1, z) \). Moreover, for any \( x \in \Omega_{\epsilon} \), we have \( x_0 + x_1 + \ldots + x_{n-1} \leq 1 \), and \( x_j \geq 0 \), thus the equation \( \varepsilon_2 y_2' = g(x, y_2, w) \) has exactly one positive steady state which is globally asymptotically stable in \( \Omega_{\epsilon} \), uniformly with respect to \( (x, w) = (x_0, \ldots, x_{n-1}, y_1, z) \) as the coefficient by \( y_2 \), that is \( -1 - \theta \sum_{j=0}^{n-1} k_j x_j \), is negative and separated from zero for all \( x \in \Omega_{\epsilon} \).

Next, we show stability of the system that describes the dynamics of number of free and occupied binding sites for the fixed concentration of dimers.

Proposition 11. Let \( f \) be the function defined by formula (3.5). For any fixed \( y_2 \in \Omega_{\epsilon} \), there exists exactly one non-negative steady state of equation \( \varepsilon_1 x' = f(x, y_2) \) which is globally asymptotically stable in \( \Omega_{\epsilon} \times \Omega_{\epsilon} \) (uniformly with respect to \( y_2 \)).
Proof. We note that the function \( f \) is linear with respect to \( x \), when \( y_2 \) is fixed. Thus, in order to determine the stability of the steady state it is enough to study the Jacobi matrix of the right hand side. Moreover, a unique steady state exists if and only if this matrix is non-singular. The existence of the non-negative steady state was already proved in Subsection 2.3.2. We prove that all eigenvalues of the matrix of the right hand side of \( \varepsilon_i \dot{x}' = f(x, y_2) \) are real and negative. To this end, we introduce additional variable and show that after the modification, the matrix becomes tridiagonal, and therefore, we use properties of tridiagonal matrices. Let us consider

\[
x_n = 1 - (x_0 + x_1 + x_2 + \cdots + x_{n-1}),
\]

and let \( \hat{x} = (x, x_n) = (x_0, x_1, \ldots, x_{n-1}, x_n) \). The variable \( \hat{x} \) fulfills the following linear ODE \( \dot{x}' = A \hat{x} + b \), where

\[
A = \begin{bmatrix}
-\kappa_{01} & \gamma_1 & 0 & 0 & \cdots & 0 \\
\kappa_{12} & -\kappa_{11} - \gamma_1 & \gamma_2 & 0 & \cdots & 0 \\
0 & \kappa_{12} & -\kappa_{22} - \gamma_2 & \gamma_3 & \cdots & 0 \\
\vdots & \ddots & \ddots & \ddots & \ddots & \vdots \\
0 & \cdots & \kappa_{n-2n-1} & -\kappa_{n-1n-1} - \gamma_{n-1} & \kappa_{n-1n-2} & \gamma_{n} \\
0 & \cdots & \cdots & 0 & \kappa_{n-1n-2} & -\gamma_{n}
\end{bmatrix},
\]

\( \hat{\gamma}_1 = j\gamma_j \), and the form of the vector \( b \) is not important here. Now, the matrix \( A \) is tridiagonal. We show that the eigenvalues of \( A \) are real and the largest of them is equal to 0. Because the studied system is linear (with respect to \( y_2 \)) and \( x_0 + x_1 + \cdots + x_{n-1} + x_n = 1 \) is an invariant subspace for the system \( \dot{x}' = A \hat{x} + b \) we deduce that the steady state of \( \varepsilon_i \dot{x}' = f(x, y_2) \) is stable.

We observe that the tridiagonal matrix \( A \) is similar to the following symmetric matrix

\[
P = \begin{bmatrix}
\alpha_1 & \beta_1 & 0 & 0 & \cdots & 0 \\
\beta_1 & \alpha_2 & \beta_2 & 0 & \cdots & 0 \\
0 & \beta_2 & \alpha_3 & \beta_3 & \cdots & 0 \\
\vdots & \ddots & \ddots & \ddots & \ddots & \vdots \\
0 & \cdots & \beta_{n-1} & \alpha_n & \beta_n \\
0 & \cdots & 0 & \beta_n & \alpha_{n+1}
\end{bmatrix},
\]

where \( \alpha_j = -\kappa_{j-1j} - \gamma_{j-1} \) are terms on the diagonal for \( 1 \leq j \leq n + 1 \) and \( \beta_j = \sqrt{k_{j-1j}\gamma_{j+1}} \) for \( 1 \leq j \leq n \) are terms on the super- and sub-diagonal. To shorten the notation, we set \( k_n = 0 \) and \( \gamma_0 = 0 \). This matrix \( P \) has a form \( D_n^{\prime} A D_n \), where \( D_n \) is a diagonal matrix with diagonal elements

\[
\delta_1 = 1, \quad \delta_j^2 = \frac{k_0 \cdots k_{j-2}}{\gamma_1 \cdots \gamma_{j-1}} y_{j-1}^{j-1}, \quad j = 2, \ldots, n.
\]

The characteristic polynomial \( \Delta_n(\lambda) = \det(P - \lambda I_n) \) can be computed by the following recurrence relations

\[
\Delta_0(\lambda) = 1, \quad \Delta_1(\lambda) = \alpha_1 - \lambda, \quad \Delta_j(\lambda) = (\alpha_j - \lambda) \Delta_{j-1}(\lambda) - \beta_j^2 \Delta_{j-2}(\lambda) \quad \text{for} \quad j \geq 2.
\]

Since similar matrices have the same eigenvalues, it is enough to examine that the eigenvalues of \( P \) are real and the largest of it is equal to 0. Observe that all eigenvalues of \( P \) are real and simple, because the matrix \( P \) is irreducible, tridiagonal and symmetric (see [14] Prop. 10.1.2). Moreover, the general theory of tridiagonal matrices implies also (see [18] Thm 5.9 (The Sturm sequence property)) that the number of eigenvalues greater than some real number \( a \) is equal to the number of agreement of sign between consecutive members of the sign sequence \( \{\Delta_0(a), \Delta_1(a), \ldots, \Delta_n(a)\} \). It is clear that \( \lambda = 0 \) is an eigenvalues of the matrix \( A \) and thus it is an eigenvalue of the matrix \( P \). Now, we prove that all eigenvalues of the members of the sequence \( \{\Delta_0(0), \Delta_1(0), \ldots, \Delta_n(0)\} \) have different signs. To this end, we prove by mathematical induction on \( n \) that

\[
\Delta_0(0) = 1, \quad \Delta_n(0) = (-1)^n k_0 \cdots k_{n-1} y_2^n \quad \text{for} \quad n \geq 1.
\]

If \( n = 1 \) this statement is obviously true, as \( \Delta_1(0) = -\kappa_0 y_2 \). Now, we show that if the formula is correct for \( \Delta_n(0) \) it is also correct for \( \Delta_{n+1}(0) \). A direct calculation shows that \( \Delta_1(0) = \kappa_0 y_2^2 \), and

\[
\Delta_{n+1} = \alpha_{n+1} \Delta_n - \beta_n^2 \Delta_{n-1} = (-\kappa_0 y_2 - \gamma_0)(-1)^n k_0 \cdots k_{n-1} y_2^n - k_{n-1} \gamma_0 y_2 (1)^{n-1} k_0 \cdots k_{n-2} y_2^{n-1} = (-1)^{n+1} k_0 \cdots k_{n-1} y_2^{n+1}.
\]

This equality ensures that successive terms of the sequence \( \{\Delta_n(0)\}_{n=0}^{\infty} \) change their sign. Thus, the matrix \( A \) has no eigenvalue greater than 0, so the system \( (x_0, x_1, \ldots, x_n) \) is stable (for fixed \( y_2 \) the system is linear). In addition, as the system is linear the stability is uniform. For this purpose, it is enough to take a diminished set \( \bar{\Omega} \), where \( y_2 \) is separated from zero (see Fig. 3). \( \square \)
As mentioned before, there are two different ways to reduce system (2.6) (or equivalently (3.3)) depending on the time scales, that is whether $\varepsilon_2 \ll \varepsilon_1 \ll 1$ (the left part of Fig. 1) or $\varepsilon_1 \ll \varepsilon_2 \ll 1$ (the right part of Fig. 1). For better readability, we present a diagram (Fig. 4) for our system (3.3).

Figure 4: Diagram of models created from the system (3.3) and connection between them.

3.2. Application of The Tikhonov theorem

Now we assume that the dimer creation is much faster process than other processes, so we assume that $\varepsilon_2$ is very small and we formulate the Tikhonov type theorem for this case. If we set $\varepsilon_2 = 0$, then system (3.3)–(3.4) reduces to system (2.14), which has the form

$$
\begin{align*}
\varepsilon_1 \dot{x} &= f(\bar{x}, y_2), \\
\dot{\bar{w}} &= h(\bar{x}, y_2, w), \quad w = (y_1, z)
\end{align*}
$$

(3.8)

where $\varphi(x, w) = y_2$ is the solution of the equation $g(x, y_2, w) = 0$, the function $\varphi$ is given by (2.13) and $f$, $g$, $h$ are defined by (3.5), (3.6), (3.7), respectively.

**Theorem 12.** Assume that the functions $f: \Omega_1 \times \Omega_2 \to \mathbb{R}$ and $g, h: \Omega_1 \times \Omega_2 \times \Omega_w \to \mathbb{R}$ are defined by (3.5)–(3.7). Then there exists $\varepsilon_0 > 0$, such that for any $\varepsilon_2 \in (0, \varepsilon_0]$ there exists a unique solution $(x_{\varepsilon_2}(t), y_{2\varepsilon_2}(t), w_{\varepsilon_2}(t))$ to system (3.3)–(3.4) on $[0, T]$ and the following conditions hold:

- $\lim_{\varepsilon_2 \to 0} x_{\varepsilon_2}(t) = \bar{x}(t)$, \quad $t \in [0, T]$,
- $\lim_{\varepsilon_2 \to 0} y_{2\varepsilon_2}(t) = \varphi(\bar{x}(t), \bar{w}(t))$, \quad $t \in (0, T]$,
- $\lim_{\varepsilon_2 \to 0} w_{\varepsilon_2}(t) = \bar{w}(t)$, \quad $t \in [0, T]$,

where $(\bar{x}(t), \bar{w}(t))$ is the solution to system (3.8) and constant $T$ does not depend on $\varepsilon_2$.

**Proof.** It is enough to check the assumptions (A1)–(A5) of Theorem 7. The assumptions (A1), and (A2) are satisfied in consequence of the form of functions $f$, $g$, and $h$. Assumption (A3) also holds as a consequence of Proposition 10. The existence of unique solution to system (3.8) is obvious due to the form of the right-hand side of the system. Note also that by Remark 7 the set $\Omega_1 \times \Omega_w \times \Omega_w$ is a trapping region for system (3.3), and the set $U = \Omega_1 \times \Omega_w$ is a trapping region for system (3.8). In the consequence the solution $(\bar{x}, \bar{w}) \in \text{Int } U$, thus assumption (A4) holds. Finally, Assumption (A5) is satisfied due to Proposition 10 and applying the Tikhonov theorem completes the proof. □
Now, we assume that the creation and dissociation of Hes1 dimers are much faster than translation and transcription processes, that is \( \varepsilon_1 \ll 1 \). For \( \varepsilon_1 = 0 \) system (3.8) reduces to the classical Hes1 model (2.19), which we rewrite as
\[
w' = h(\psi(w), \varphi(\psi(w), w)), \quad w(0) = \dot{w},
\]
where \( \psi(w) = x \) is the solution of the equation \( f(x, \varphi(x, w)) = 0 \), and the function \( \psi \) is given by (2.17). Now, we state theorem saying that for \( \varepsilon_1 \) small enough, the solutions to system (3.9) approximate solutions to system (3.8) well.

**Theorem 13.** Assume that the functions \( f: \Omega_x \times \Omega_z \rightarrow \mathbb{R} \) and \( h: \Omega_x \times \Omega_z \times \Omega_w \rightarrow \mathbb{R} \) are defined by (3.5) and (3.7), respectively. Then there exists \( \varepsilon_0 > 0 \), such that for any \( \varepsilon_1 \in (0, \varepsilon_0] \) there exists a unique solution \( (x_{\varepsilon_1}(t), w_{\varepsilon_1}(t)) \) to system (3.8) on \([0, T] \) and the following conditions hold:
\[
\lim_{\varepsilon_1 \to 0} x_{\varepsilon_1}(t) = \psi(\tilde{w}(t)), \quad t \in (0, T]
\]
\[
\lim_{\varepsilon_1 \to 0} w_{\varepsilon_1}(t) = \tilde{w}(t), \quad t \in [0, T]
\]
where \( \tilde{w}(t) \) is the solution to system (3.9) and constant \( T \) is independent of \( \varepsilon_1 \).

Proof. It is easy to see that smoothness assumptions (A1) and (A2) of the Tikhonov theorem are satisfied. The stability properties, (A3) and (A5), as well as assumption (A4) hold, due to analogous argument as in the proof of Theorem 12 when we use Proposition 11 instead of Proposition 10. Thus, Theorem 7 yields the assertion of Theorem 13. \( \square \)

Assuming that the creation and dissociation of Hes1 dimers are much faster than translation and transcription processes, that is \( \varepsilon_1 \) is very small we reduce system (3.3)–(3.4) to system (2.18), which reads
\[
\begin{align*}
\varepsilon_2 \dot{y}_2 &= g(\varphi(y_2), \tilde{y}_2, \tilde{w}), \\
\dot{w}' &= h(\varphi(y_2), \tilde{y}_2, \tilde{w}),
\end{align*}
\]
where \( \varphi(y_2) = x \) is the solution of the equation \( f(x, y_2) = 0 \), and the function \( \varphi \) is given by (2.13).

**Theorem 14.** Assume that the functions \( f: \Omega_x \times \Omega_y \rightarrow \mathbb{R} \) and \( g: \Omega_x \times \Omega_y \times \Omega_w \rightarrow \mathbb{R} \) are defined by (3.5)–(3.7). Then there exists \( \varepsilon_0 > 0 \), such that for any \( \varepsilon_1 \in (0, \varepsilon_0] \) there exists a unique solution \( (x_{\varepsilon_1}(t), y_{\varepsilon_1}(t), w_{\varepsilon_1}(t)) \) to system (3.3)–(3.4) on \([0, T] \) and the following conditions hold:
\[
\lim_{\varepsilon_1 \to 0} x_{\varepsilon_1}(t) = \varphi(\tilde{w}(t)), \quad t \in (0, T]
\]
\[
\lim_{\varepsilon_1 \to 0} y_{\varepsilon_1}(t) = \tilde{y}(t), \quad t \in [0, T]
\]
\[
\lim_{\varepsilon_1 \to 0} w_{\varepsilon_1}(t) = \tilde{w}(t), \quad t \in [0, T],
\]
where \( (\tilde{y}(t), \tilde{w}(t)) \) is the solution to system (3.10) and constant \( T \) is independent of \( \varepsilon_1 \).

In the next step we put \( \varepsilon_2 \approx 0 \) and we reduce system (3.10) to the classical form
\[
\dot{w}' = h(\varphi(\psi(\dot{w})), \varphi(\dot{w})), \quad \dot{w}(0) = \dot{w},
\]
where \( \psi(w) = y_2 \) is the solution of the equation \( g(\varphi(y_2), \tilde{y}_2, \tilde{w}) = 0 \).

**Theorem 15.** Assume that the functions \( g, h: \Omega_x \times \Omega_y \times \Omega_w \rightarrow \mathbb{R} \) are defined by (3.6)–(3.7). Then there exists \( \varepsilon_0 > 0 \), such that for any \( \varepsilon_2 \in (0, \varepsilon_0] \) there exists a unique solution \( (x_{\varepsilon_2}(t), w_{\varepsilon_2}(t)) \) to system (3.8) on \([0, T] \) and the following conditions hold:
\[
\lim_{\varepsilon_2 \to 0} x_{\varepsilon_2}(t) = \psi(\tilde{w}(t)), \quad t \in (0, T]
\]
\[
\lim_{\varepsilon_2 \to 0} w_{\varepsilon_2}(t) = \tilde{w}(t), \quad t \in [0, T]
\]
where \( \tilde{w}(t) \) is the solution to system (3.11) and constant \( T \) does not depend on \( \varepsilon_2 \).

The proofs of the Theorems 14 and 15 are analogously as Theorems 12 and 13. It is worth pointing out that the crucial assumption of these theorems is the Assumption (A3) of the Tikhonov theorem, i.e. the steady state of the respective systems is asymptotically stable independently of the parameters value, which is guaranteed by Propositions 11 and 10.
4. Comparison of stability of the positive steady state

In this section we examine stability of the steady states for systems derived in Section 2. The first two subsections deal with the case of one binding site (i.e. \( n = 1 \)) because with the increase of \( n \) the complexity of the equations makes their analysis difficult.

4.1. The full system — the case \( n = 1 \)

Note that due to the scaling we have \( k_0 = 1 \). In (2.6), we wrote explicitly \( k_0 \) due to the symmetry of the notation. However, here we rewrite system (2.6) for \( n = 1 \) and \( k_j \) for \( j > 0 \) are not present. Thus, we use the equality \( k_0 = 1 \) replacing \( k_0 \) by 1, and the system reads

\[
\begin{align*}
\varepsilon_1 x_0' &= \gamma_1 (1 - x_0) - y_2 x_0, \\
y_1' &= k(y_2 - y_1^2) + \delta_1 (z - y_1), \\
\varepsilon_2 y_2' &= \theta(y_1 (1 - x_0) - x_0 y_2) - y_2 + y_1^2, \\
\varepsilon' &= \delta_2 (n_0 x_0 - z).
\end{align*}
\]

(4.1)

Theorem 16. The positive steady state \((\bar{x}_0, 1, 1)\) of system (4.1) is locally asymptotically stable.

Proof. In order to calculate stability of the positive steady state \((\bar{x}_0, 1, 1)\) of system (4.1), where \( \bar{x}_0 = 1/r_0, \ r_0 = 1 + 1/\gamma_1 \), we rescale time to eliminate \( \varepsilon_2 \) from the left-hand side and then the Jacobi matrix reads

\[
\begin{bmatrix}
-(\gamma_1 + 1) & 0 & -\bar{x}_0 \varepsilon & 0 \\
0 & -(2k + \delta_1) & \varepsilon_2 - \lambda & k \varepsilon_2 \\
-\theta(\gamma_1 + 1) & 2 & -((\theta \bar{x}_0 + 1) - \lambda) & 0 \\
\varepsilon_2 \bar{x}_0 \varepsilon & 0 & 0 & -\varepsilon_2 \delta_2 - \lambda
\end{bmatrix},
\]

where \( \varepsilon = \frac{r_0}{\bar{x}_1} \). The characteristic polynomial in this case reads

\[
W_1(\lambda) = \lambda^4 + a_1 \lambda^3 + a_2 \lambda^2 + a_3 \lambda + a_4,
\]

(4.2)

where

\[
a_1 = \varepsilon \varepsilon_2 \delta_2 + c_2, \quad a_2 = \varepsilon \varepsilon_2 \delta_2 c_2 + c_1, \quad a_3 = \varepsilon \varepsilon_2 \delta_2 c_1 + c_0, \quad a_4 = \varepsilon \varepsilon_2 \delta_2 c_0 + 2 \varepsilon \varepsilon_2 \delta_1 \delta_2
\]

(4.3)

(we used the fact that \( r_0 \bar{x}_0 = 1 \)), and

\[
\begin{align*}
c_0 &= \varepsilon \varepsilon_2 \delta_1 (\gamma_1 + 1) > 0, \\
c_1 &= \varepsilon (\gamma_1 + 1) (\varepsilon_2 (2k + \delta_1) + 1) + \varepsilon_2 (2 \theta \bar{x}_0 + \delta_1 (1 + \theta \bar{x}_0)) > 0, \\
c_2 &= \varepsilon (\gamma_1 + 1) + \varepsilon_2 (2k + \delta_1) + \theta \bar{x}_0 + 1 > 0.
\end{align*}
\]

We check that all eigenvalues of the characteristic polynomial \( W_1(\lambda) \) are negative. We observe that all coefficients \( a_i \) of (4.2) are positive. Then according to the Hurwitz theorem, it is enough to show that

\[
a_1 a_2 a_3 > a_1^2 + a_1^2 a_4,
\]

which is equivalent to \( a_3 (a_1 a_2 - a_3) > a_1^2 a_4 \). Note that the necessary condition \( a_1 a_2 - a_3 > 0 \) occurs

\[
a_1 a_2 - a_3 = (\varepsilon \varepsilon_2 \delta_2 + c_2) \varepsilon \varepsilon_2 \delta_2 c_2 + c_1 c_2 - c_0 = \varepsilon \varepsilon_2 \delta_2 c_2 (\varepsilon \varepsilon_2 \delta_2 + c_2) + c_3 (c_0 + c_4) + c_4 > 0,
\]

where

\[
c_3 = \varepsilon (\gamma_1 + 1) + \varepsilon_2 (2k + \delta_1) + \theta \bar{x}_0, \quad c_4 = \varepsilon (\gamma_1 + 1) (2 \varepsilon_2 k + 1) + \varepsilon_2 (2 \theta \bar{x}_0 + \delta_1 (1 + \theta \bar{x}_0)).
\]

Note also that

\[
a_4 = \varepsilon \varepsilon_2 \delta_1 \delta_2 (\gamma_1 + 3) > 0.
\]

Using the formulas for the coefficients \( a_i \), i.e. (4.3), after tedious but direct calculations, we obtain \( a_3 (a_1 a_2 - a_3) - a_1^2 a_4 > 0 \). We include these computations in the Appendix. \( \Box \)
4. Model without dimers — the case $n = 1$

For $n = 1$ formula (2.13) takes the form

$$y_2 = \frac{y_1^2 + \theta y_1 (1 - x_0)}{1 + \theta x_0}$$

and using the equality $x_0 + x_1 = 1$ we write system (2.18) for $n = 1$ as

$$\begin{align*}
\varepsilon_1 x_0' &= \frac{1}{1 + \theta x_0} \left( \gamma_1 (1 - x_0) - x_0 y_1^2 \right), \\
y_1' &= \frac{k \theta}{1 + \theta x_0}\left( \gamma_1 (1 - x_0) - x_0 y_1^2 \right) + \delta_1 (z - y_1), \\
z' &= \delta_2 (r_0 x_0 - z).
\end{align*}$$

(4.4)

We observe that the steady state is

$$\left( \frac{\gamma_1}{1 + \gamma_1}, 1, 1 \right).$$

**Theorem 17.** There exists a unique positive steady state of system (4.4), which is asymptotically stable independently of the parameters value.

**Proof.** The characteristic matrix for the steady state $\left( \frac{\gamma_1}{1 + \gamma_1}, 1, 1 \right)$ of (4.4) reads

$$\begin{bmatrix}
-\frac{\eta(1+\gamma_1)}{e_1} & -\frac{2a y_1}{e_1^2} & 0 \\
-k \eta \theta (1 + \gamma_1) & -\frac{a y_1}{e_1^2} - \delta_1 & \delta_1 \\
\delta_2 & \frac{\gamma_1}{1 + \gamma_1} & 0 - \delta_2
\end{bmatrix}, \quad \eta = \frac{1 + \gamma_1}{1 + \gamma_1 (1 + \theta)}.$$

Then the characteristic function takes the form

$$W(\lambda) = \lambda^3 + a_1 \lambda^2 + a_2 \lambda + a_3,$$

where

$$\begin{align*}
a_1 &= \delta_1 + \delta_2 + \eta \frac{1 + \gamma_1}{e_1} + 2 k \eta \theta \frac{\gamma_1}{1 + \gamma_1}, \\
a_2 &= \eta (\delta_1 + \delta_2) \frac{1 + \gamma_1}{e_1} + 2 k \eta \theta \delta_2 \frac{\gamma_1}{1 + \gamma_1} + \delta_1 \delta_2, \\
a_3 &= \frac{\eta \delta_1 \delta_2}{e_1} (3 + \gamma_1).
\end{align*}$$

(4.5)

For the polynomial $W(\lambda)$ of degree 3 the Routh-Hurwitz criterion simplify to $a_1, a_3 > 0$ and $a_1 a_2 > a_3$. Note that $a_1, a_3 > 0$ and the condition $a_1 a_2 > a_3$ is equivalent to

$$\eta \left( \delta_1 + \delta_2 + \frac{1 + \gamma_1}{e_1} + 2 k \eta \theta \frac{\gamma_1}{1 + \gamma_1} \right) \left( \eta (\delta_1 + \delta_2) \frac{1 + \gamma_1}{e_1} + 2 k \eta \theta \delta_2 \frac{\gamma_1}{1 + \gamma_1} + \delta_1 \delta_2 \right) > \frac{\eta \delta_1 \delta_2}{e_1} (3 + \gamma_1).$$

Moreover, the above inequality is always fulfilled. Namely, multiplication of $\delta_1 + \delta_2$ from the first bracket by the first term of the second bracket gives $\frac{\eta}{e_1} (\delta_1 + \delta_2)^2 (1 + \gamma_1)$. Adding to this expression the second term of the first bracket multiplied by the last term of the second bracket gives

$$\eta \left( \delta_1 + \delta_2 \right)^2 (1 + \gamma_1) + \frac{\eta \delta_1 \delta_2}{e_1} (1 + \gamma_1) = \frac{3 \eta \delta_1 \delta_2}{e_1} (1 + \gamma_1) + \eta \frac{(\delta_1^2 + \delta_2^2) (1 + \gamma_1)}{e_1} > \frac{\eta \delta_1 \delta_2}{e_1} (3 + \gamma_1).$$

This completes the proof. □
4.3. Model with dimers

Now, we formulate theorem considering stability of the steady state (1, 1, 1) of the system (2.18).

**Theorem 18.** If the function \( \psi \) given by (2.17) satisfies the following inequality
\[
-\psi'(1) < \frac{\varepsilon_2(2k + \delta_1) + 1}{2k_0\varepsilon_2} \left( \frac{\varepsilon_2\delta_2(2k + \delta_1 + \delta_2) + \delta_1 + \delta_2}{2k_0\varepsilon_2} \right),
\]
then the steady state (1, 1, 1) of system (2.18) is locally asymptotically stable.

**Proof.** Linearising the system (2.18) around the steady state (1, 1, 1) we obtain the following matrix
\[
\begin{pmatrix}
-A & k & \delta_1 \\
2/\varepsilon_2 & -1/\varepsilon_2 & 0 \\
0 & r_0\delta_2\psi'(1) & -\delta_2
\end{pmatrix}, \quad A = 2k + \delta_1.
\]
The characteristic polynomial reads
\[
W(\lambda) = \lambda^3 + \left(A + \delta_2 + \frac{1}{\varepsilon_2}\right)\lambda^2 + \left(\frac{\delta_1}{\varepsilon_2} + \delta_2\left(A + \frac{1}{\varepsilon_2}\right)\right)\lambda + \frac{\delta_1\delta_2}{\varepsilon_2}(1 - 2r_0\psi'(1)).
\]
Due to positivity of the coefficient and negativity of \( \psi'(1) \) we can immediately conclude that all coefficients of polynomial \( W \) are positive. Thus, due to the Routh-Hurwitz criterion, the steady state is locally stable if the inequality
\[
\left(A + \delta_2 + \frac{1}{\varepsilon_2}\right)\left(\frac{\delta_1}{\varepsilon_2} + \delta_2\left(\frac{1 + 1}{\varepsilon_2}\right)\right) > \frac{\delta_1\delta_2}{\varepsilon_2}(1 - 2r_0\psi'(1))
\]
which is equivalent to
\[
\left(A\varepsilon_2 + 1\right)\left(\delta_1 + \delta_2(A\varepsilon_2 + 1) + \varepsilon_2\delta_2\right) > -2r_0\delta_1\delta_2\varepsilon_2\psi'(1).
\]
This completes the proof. \( \Box \)

**Remark 19.** If the strict inequality reverse to (4.6) holds, that is
\[
-\psi'(1) > \frac{\varepsilon_2(2k + \delta_1) + 1}{2k_0\varepsilon_2} \left( \frac{\varepsilon_2\delta_2(2k + \delta_1 + \delta_2) + \delta_1 + \delta_2}{2k_0\varepsilon_2} \right),
\]
then the steady state (1, 1, 1) of system (2.18) is unstable.

Recall that due to the chosen scaling equality \( \psi(1) = \frac{1}{r_0} \) holds, where \( r_0 \) is defined by (2.5), i.e.
\[
r_0 = 1 + \sum_{j=1}^{n} \frac{1}{\gamma_j} - \frac{1}{\gamma_j} - \frac{1}{\gamma_j}.
\]
The stability condition presented in Theorem 18 is valid for an arbitrary \( C^1 \) class function \( \psi \). However, due to the origin of system (2.18), the function \( \psi \) has a specific form given by (2.17). Therefore, both sides of inequality (4.6) depend on the number of binding sites \( n \). In the following, we prove that if the number of binding sites is small (not greater than 4) then the steady state is always stable, regardless of the value of other parameters. The destabilisation of the steady state is possible if there is at least 5 binding sites.

**Theorem 20.** If the function \( \psi \) is given by (2.17), then

1. for \( n \leq 4 \) the steady state of system (2.18) is locally asymptotically stable,
2. for \( n \geq 5 \) there exists set of parameters of model (2.18) as well as the function \( \psi \) such that the steady state of system (2.18) is unstable. More precisely, if \( n \geq 5, \delta_1, \delta_2 \) and \( \varepsilon_2 \) sufficiently close to 1, \( k \) small enough, \( \psi(y) = \frac{1}{1 + (r_0 - 1)y^2} \) and \( r_0 > \frac{n}{n - 4} \), then the steady state of system (2.18) is unstable.
Lemma 21. Let \( \psi(y) = \frac{1}{1+q(y)} \), where \( q \) is a polynomial of \( n \)-th degree with all coefficient non-negative such that \( q(0) = 0 \). Then

\[
-\psi'(1) \leq \frac{\nu q(1)}{(1 + q(1))^2}.
\]

Proof. Let us fix the value \( q(1) \). We obtain that

\[
\psi'(1) = -\frac{q'(1)}{(1 + q(1))^2}.
\]

We need to find an upper bound of the numerator of \(|\psi'(1)|\). Let \( q(y) = \sum_{j=1}^n a_j y^j \). Then this numerator reads \( \sum_{j=1}^n j a_j \).

Since all coefficients \( a_j \) of \( q \) are non-negative, we have

\[
q'(1) = \sum_{j=1}^n j a_j = n q(1) - \sum_{j=1}^{n-1} (n - j) a_j \leq n q(1),
\]

which proves our assertion. \( \square \)

Proof of Theorem 20. Let us prove the first part of our theorem. By (2.5) and (2.17) we get \( \psi(1) = 1/r_0 \). Using notation of Lemma 21 we get \( r_0 = 1 + q(1) \). Thus, due to Lemma 21 we obtain

\[
-\psi'(1) \leq \frac{n(r_0 - 1)}{r_0^2}.
\]

Suppose that the steady state is not locally asymptotically stable. Then, the inequality reverse to (4.6) holds and we get

\[
\frac{n(r_0 - 1)}{r_0^2} \geq -\psi'(1) \geq \frac{\nu (2 \hat{\delta}_2 + \hat{\delta}_1 + 1)(2 \hat{\delta}_2 + \hat{\delta}_1 + \hat{\delta}_2 + \hat{\delta}_2)}{2 \nu \hat{\delta}_2 \hat{\delta}_1 \hat{\delta}_2}.
\]

Thus,

\[
n \geq \frac{r_0}{r_0 - 1} \frac{(2 \hat{\delta}_2 + \hat{\delta}_1 + 1)(2 \hat{\delta}_2 + \hat{\delta}_1 + \hat{\delta}_2 + \hat{\delta}_2)}{2 \nu \hat{\delta}_2 \hat{\delta}_1 \hat{\delta}_2}.
\]

Observe, that as \( k > 0 \), the second term of (4.8) can be estimated as

\[
\frac{(2 \hat{\delta}_2 + \hat{\delta}_1 + 1)(2 \hat{\delta}_2 + \hat{\delta}_1 + \hat{\delta}_2 + \hat{\delta}_2)}{2 \nu \hat{\delta}_2 \hat{\delta}_1 \hat{\delta}_2} \geq \frac{(\hat{\delta}_1 + \hat{\delta}_2)^2}{2 \nu \hat{\delta}_2 \hat{\delta}_1} + \frac{1}{2} \left( \frac{1}{\hat{\delta}_2 \hat{\delta}_1} + \frac{\hat{\delta}_1}{\hat{\delta}_2} \right) \geq 4.
\]

Combining (4.8) with (4.9) we obtain

\[
n \geq \frac{4 r_0}{r_0 - 1} > 4,
\]

which contradicts our assumption.

To prove the second part of our theorem we observe that inequalities (4.9) becomes equalities for \( \hat{\delta}_2 = \hat{\delta}_1 = \hat{\delta}_2 = 1 \) and \( k = 0 \). Thus, for \( n \geq 5 \), for \( r_0 \) sufficiently large, that is greater than

\[
r_0 > \frac{n}{n - 4}
\]

and for sufficiently small \( k \) inequality (4.8) holds. As inequality proved in Lemma 21 becomes equality for \( q(y) = a y^n \), inequality (4.7) holds for \( n \geq 5 \), \( k \) sufficiently small, and \( \hat{\delta}_1, \hat{\delta}_2 \) sufficiently close to 1. This completes the proof. \( \square \)

5. Discussion and conclusions

The theorems stated in the previous sections show, that the dynamics of the reduced models (a slow part of the model) is similar to the model before reduction (that contains both parts: slow and fast) if the „fast” subsystem is sufficiently fast (that is \( \varepsilon_1 \) or \( \varepsilon_2 \) is sufficiently small). On the other hand, we have also showed that the dynamics of such model can differ from the dynamics of the reduced system. In particular, in the classical Hes1 system (2.19) without time delay the positive steady state is always (globally) asymptotically stable. However, if we take into account Hes1 dimers as a separate population, that is system (2.18), we can see that the steady state can loose stability and oscillatory behaviour is possible (if the number of binding sites is greater than 4). Now, we numerically illustrate similarities and differences in behaviour of solutions to the considered models.
5 DISCUSSION AND CONCLUSIONS

5.1. Numerical simulations

In order to show behaviour of four models considered in this paper (full, without dimers, with dimers and classical), basing on data from \[11\], we choose the following parameters

\[
\delta_1 = 0.2242, \quad \delta_2 = 0.2075, \quad \theta = 0.5 \quad \text{and} \quad \epsilon_1 = 1. \quad (5.1)
\]

Our models, before scaling, consist a vast number of parameter. Estimating those parameter may be a challenging problem that we are not going to address here. Our aim is only to illustrate possible models’ behaviours. Thus, we choose the parameter values only for the non-dimensional version of our models ensuring that the ratio between \(\delta_1\) and \(\delta_2\) agrees with protein decay and mRNA decay rates considered in \[11\]. We illustrate the behaviour of considered models for two different situations. In the first case, we consider the situation, when the steady states of all models are stable and oscillatory behaviour is not possible. We put

\[
k = 0.2, \quad \epsilon_2 = 1, \quad k_0 = 1, \quad k_1 = 1.5, \quad k_2 = 1.5, \quad \gamma_1 = 1, \quad \gamma_2 = 0.5, \quad \gamma_3 = 0.5. \quad (5.2)
\]

Here, we take \(\gamma_i\) and \(k_i\) to reflect the cooperative character of the binding of dimer to DNA promoter. Thus, binding ratio is larger and the dissociation parameter is smaller if at least one binding site is occupied. The result of an exemplary simulation is presented in Fig. 5. Although the difference in solutions to different models is visible, the qualitative behaviour of those solutions are the same. In particular, all solutions converge to the stationary state.

![Figure 5: Comparison of solutions to system considered in the paper for the case \(n = 3\). Parameters as given in (5.1) and (5.2).](image)

On the other hand, we also want to illustrate possible oscillatory behaviour of system with dimers \(5.18\) if the derivative of function \(\psi\) (given by \(2.17\)) at the stationary state is large enough. In order to do that we need to take \(n \geq 5\). We decided to take the smallest possible \(n = 5\), but then we need to take \(k\) small enough and appropriate \(k_i\) and \(\gamma_i\) such that the inequality reversed \(4.6\) holds. We take

\[
k = 0.01, \quad \epsilon_2 = 5, \quad k_0 = 1, \quad k_1 = 2, \quad k_2 = 3, \quad k_3 = 4, \quad k_4 = 700, \quad \gamma_1 = 2, \quad \gamma_i = 1, \quad i = 2, \ldots, 5. \quad (5.3)
\]

The results of the simulations are presented in Fig. 6. The interesting fact is that only solutions to \(2.18\) exhibit oscillatory behaviour. Of course, if we take \(\epsilon_1\) sufficiently close to zero, the oscillatory behaviour appears also for the full model \(2.6\).

Note that we need to take \(k_4\) very large in order to fulfil condition \(4.7\) for \(n = 5\) and \(\delta_1, \delta_2\) as in \(5.1\). This causes that we need very small \(\epsilon_1\) to get an agreement in behaviour of solutions to the model with dimers and to the full model. However, if we consider larger number of binding sites, we may take \(k_i\) of order 1. For example, for \(n = 9\) of binding sites and \(k_i = i + 1\) we observe oscillatory behaviour of the solution to the model with dimers and dumping oscillation in the full model for \(\epsilon_1 = 1\) (compare the upper row of Fig. 7) or oscillation to both models for \(\epsilon_1 = 0.05\) (compare the bottom row of Fig. 7).

5.2. Conclusions

In this paper we examine a gene expression model of the self-regulating Hes1 protein. We take into account the existence of multiple binding sites in the Hes1 promoter and the model of the transcriptional and translational processes. In our model, we include negative feedback and we take into account the formation of dimer complexes by proteins.
APPENDIX A  THE CALCULATION NEEDED IN THE PROOF OF STABILITY OF THE STEADY STATE OF SYSTEM (4.1)

Here we include the calculations that proves the stability of the steady state of system (4.1). The notation is as in Section 4.1, in particular $a_i$ are defined by (4.3).

$$
a_3(a_1a_2 - a_3) - a_1^2a_4 =$$

$$
e_2(e_2^2(\chi_2^2e_2^2 + (\varepsilon^2 + (\theta_0 + \varepsilon_1)\chi + \varepsilon + 1) + 1)\delta_2e_2 + e_2(\gamma_1 + 1))\delta_2^2$$

$$+ e_2(\delta_2^2(2k(3\chi - 1) + \chi\delta_2)\chi_2^3 + 2\delta_2(k(3\varepsilon^2 + (\theta_0 + \varepsilon_1)(\chi + 3\varepsilon + 1) + 1) + (\varepsilon^2 + (\theta_0 + \varepsilon_1)(\chi + 3\varepsilon + 1) + 1)\delta_2)^2\chi_2^2$$

$$+ (2\varepsilon\gamma_1 + 1)(\chi - 1) + \varepsilon_3^2\chi_2^3 + 2\varepsilon_3(\gamma_1\varepsilon + 2\varepsilon_1)\chi_0 + (\varepsilon - 1)\varepsilon + \varepsilon_3(\gamma_1\varepsilon + 2\varepsilon_1) + 1)\delta_2 + \varepsilon_3(\gamma_1 + 1)\varepsilon_3^2\delta_2^2 +$$

$$+ (2k\delta_2^2(6k\chi - 4k + \delta_2(\chi - 1))\varepsilon_4^2 + 2\varepsilon_4(4\varepsilon_2 + (\theta_0 + \varepsilon_1)(3\chi + 3\varepsilon - 1))\delta_2 +$$

$$+ 2\varepsilon\gamma_1 + 1 + 2\varepsilon^2(6\varepsilon + 6\theta_0)\gamma_2^2 + 2\varepsilon(2(\varepsilon + 1)(3\varepsilon + 2) + \theta_0(12\varepsilon + 6\theta_0 + 11)\gamma_1 + 4\varepsilon^2(2\varepsilon + 3) +$$

$$+ \theta_0(6\varepsilon + 7) + 2\theta_0(3\varepsilon + \theta_0 + 2) + 2)\gamma_1$$

where

$$\zeta = \varepsilon\gamma_1 + \varepsilon + \theta_0 + 1.$$
Figure 7: Comparison of solutions to system considered in the paper for the case $n = 9$. Parameters as given in (5.1) and (5.3), except that $k_i = i + 1$, $\gamma_1 = 1$ and $\epsilon_1 = 1$ (upper row) or $\epsilon_1 = 0.05$ (lower row).

Since the following estimation holds

$$(\epsilon - 1)\epsilon + \epsilon \gamma_1 \left(\gamma_1 \epsilon + 2\epsilon + 3\right) + 1 \geq (\epsilon - 1)\epsilon + 1 = \epsilon^2 - \epsilon + 1 \geq 0$$

for all $\epsilon \geq 0$, an easy calculation shows that $a_3(a_1a_2 - a_3) - a_1^2a_4 > 0$.

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