Stability and Uniqueness of Slowly Oscillating Periodic Solutions to Wright's Equation

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Abstract

In this paper, we prove that Wright's equation $y'(t) = -\alpha y(t-1)\{1+y(t)\}$ has a unique slowly oscillating periodic solution (SOPS) for all parameter values $\alpha \in [1.9, 6.0]$, up to time translation. Our proof is based on a same strategy employed earlier by Xie [27]; show that every SOPS is asymptotically stable. We first introduce a branch and bound algorithm to control all SOPS using bounding functions at all parameter values $\alpha \in [1.9, 6.0]$. Once the bounding functions are constructed, we then control the Floquet multipliers of all possible SOPS by solving rigorously an eigenvalue problem, again using a formulation introduced by Xie. Using these two main steps, we prove that all SOPS of Wright's equation are asymptotically stable for $\alpha \in [1.9, 6.0]$, and the proof follows. This result is a step toward the proof of the Jones' Conjecture formulated in 1962.

Key words. Wright's Equation · Jones's Conjecture · Delay Differential Equations Computer-Assisted Proofs · Branch and Bound

1 Introduction

In [25] Wright studied the delay differential equation (DDE)

$$y'(t) = -\alpha y(t-1)\{1 + y(t)\}\tag{1}$$

and showed that if $\alpha > \frac{\pi}{2}$ and a solution y(t) was positive for $t \in (0,1)$, then y does not approach 0 as $t \to \infty$ and there are infinitely many zeros of y separated by a distance greater than unity. If a periodic solution has this property it is said to be "slowly oscillating" and is formally defined as follows.

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Definition 1.1. A slowly oscillating periodic solution (SOPS) is a periodic solution y(t) with the following property: there exist $q, \bar{q} > 1$ and $L = q + \bar{q}$ such that up to a time translation, y(t) > 0 on the interval (0,q), y(t) < 0 on the interval (q,L), and y(t+L) = y(t) for all t, so that L is the minimal period of y(t).

The existence of SOPS to (1) for all $\alpha > \frac{\pi}{2}$ was proven in 1962 by Jones [5] who formulated the following conjecture based on numerical experiments [6]:

Conjecture 1.2 (Jones' Conjecture). For all $\alpha > \frac{\pi}{2}$ there exists a unique SOPS to (1) (up to time translation).

We briefly describe results on the global dynamics of Wright's equation. At $\alpha = \frac{\pi}{2}$ there is a super-critical Hopf bifurcation [3]. This branch of SOPS grows without bound, that is for each L>4 this branch contains SOPS of period L [14] and for all $\alpha>\frac{\pi}{2}$ this branch contains SOPS at parameter α [13]. For $\alpha>\frac{\pi}{2}$ there is an asymptotically stable annulus in the (x(t),x(t-1)) plane whose boundary is a pair of slowly oscillating periodic orbits [7]. In [26] Xie showed that if $\alpha \geq 5.67$, then there is a unique SOPS to Wright's equation. In this paper using the computer we are able to extend Xie's method of proof thereby obtaining the following result.

Theorem 1.3. There exists a unique SOPS to Wright's equation for $\alpha \in [1.9, 6.0]$.

Combining Theorem 1.3 with the work of [26] it follows that there exists a unique SOPS to Wright's equation for $\alpha \geq 1.9$.

The set of all periodic orbits to Wright's equation for $\alpha > 0$ form a 2-dimensional manifold [16]. Based on this, it is proposed in [9] to divide Conjecture 1.2 into two parts:

- (1) there are no saddle-node bifurcations in the branch of SOPS emanating from the Hopf bifurcation at $\alpha = \frac{\pi}{2}$, and
- (2) there are no other connected components (isolas) of SOPS for $\alpha > \frac{\pi}{2}$.

Using computer-assisted proofs, it has been shown that saddle-node bifurcations do not occur for neither $\alpha \in [\frac{\pi}{2} + \epsilon_1, 2.3]$ where $\epsilon_1 = 7.3165 \times 10^{-4}$ [9] nor $\alpha \in (\frac{\pi}{2}, \frac{\pi}{2} + \epsilon_2)$ for $\epsilon_2 = 6.830 \times 10^{-3}$ [23]. Together with these results, Theorem 1.3 fills in the gap $\alpha \in (2.3, 5.67)$ needed to resolve part (1) of the Jones' conjecture (see [23, Corollary 4.8] for the complete proof). Furthermore, the paper [23] explicitly constructs a neighborhood about the bifurcation point at $\alpha = \frac{\pi}{2}$ within which there exists at most one SOPS to Wright's equation. The proof for Theorem 1.3 is given in Section 6. Thereby the Jones conjecture (Conjecture 1.2) is reduced to the following open problem:

Conjecture 1.4. There are no isolar of SOPS for $\alpha \in (\frac{\pi}{2}, 1.9)$.

To study Wright's equation, we make the change of variables $x = \ln(1 + y)$, obtaining the equivalent differential equation

$$x'(t) = -\alpha f(x(t-1)) \tag{2}$$

where $f(x) = e^x - 1$, which we will hereafter refer to as Wright's equation. Critical to Xie's result on the Jones conjecture is the relation between the asymptotic dynamics of SOPS and their global uniqueness.

Theorem 1.5 (See [26, 27]). If $\alpha > \frac{\pi}{2}$ and every SOPS to (2) is asymptotically stable, then (2) has a unique SOPS up to a time translation.

Following Xie's approach, we define the function space

$$\mathcal{X} \stackrel{\text{def}}{=} \left\{ x \in C^1(\mathbb{R}, \mathbb{R}) \mid x(0) = 0, x'(0) > 0 \text{ and } x(t) < 0 \text{ for } t \in (-1, 0) \right\}.$$

Up to a time translation, the space \mathcal{X} contains all SOPS to Wright's equation. Xie showed that if $x \in \mathcal{X}$ is a SOPS to Wright's equation with period L, then its nontrivial Floquet multipliers $\lambda \in \mathbb{C}$ are given by solutions to the nonautonomous linear DDE:

$$y'(t) = -\alpha f'(x(t-1))y(t-1)$$
(3)

subject to the boundary condition

$$\lambda y(s) = -y(L)\frac{x'(s+L)}{x'(L)} + y(s+L), \qquad s \in [-1, 0]. \tag{4}$$

For a SOPS $x \in \mathcal{X}$, showing that $|\lambda| < 1$ for all possible solutions y to (3) and (4) it suffices to show that x is asymptotically stable. By doing so for all possible SOPS to Wright's equation when $\alpha \geq 5.67$, Xie achieved his proof for uniqueness. Xie's method has two parts: (1) obtain estimates on SOPS to Wright's equation and (2) use these estimates to develop an upper bound on the magnitude of their Floquet multipliers. Xie was only able to obtain a proof for $\alpha \geq 5.67$ because of the difficulty of the first part. In this paper we continue Xie's method by means of a computer-assisted proof.

Our approach to obtaining bounds on SOPS is based on an algorithmic case-by-case analysis of the locations of the zeros of a function $x \in \mathcal{X}$ and the size of its extrema. In [25, Lemmas 4 and 5] it is shown that if $x \in \mathcal{X}$ and $\alpha > 1$ then the zeros $\{z_i(x)\}_{i=0}^{\infty}$ of x are countably infinite and $z_{i+1}(x) - z_i(x) > 1$. This result implies that we can define the maps $q : \mathcal{X} \to (1, \infty)$ and $\bar{q} : \mathcal{X} \to (1, \infty)$ as follows given $x \in \mathcal{X}$:

$$q(x) \stackrel{\text{def}}{=} z_1(x) - z_0(x),$$

 $\bar{q}(x) \stackrel{\text{def}}{=} z_2(x) - z_1(x).$

By construction, if $x \in \mathcal{X}$, then its first zero is $z_0(x) = 0$. Moreover, if x is a SOPS then $q(x) + \bar{q}(x)$ is its period and furthermore if it solves (2), then its extrema are given as

$$\begin{aligned} \max_{t \in \mathbb{R}} x(t) &= x(1), \\ \min_{t \in \mathbb{R}} x(t) &= x \left(q(x) + 1 \right). \end{aligned}$$

In [1] a branch and bound algorithm is applied to the 2-dimensional domain $\{\max x, \min x\}$ to show that there do not exist any SOPS to Wright's equation for $\alpha \leq 1.5706$, making substantial progress on Wright's conjecture that the origin is the global attractor to (1) for $\alpha < \frac{\pi}{2} \approx 1.57079$. Without an exact value for q(x), one cannot pinpoint the location of the minimum of x. To account for this ambiguity, the authors in [1] use a collection of six different functions to bound x, each defined relative to one of the zeros $\{z_0(x), z_1(x), z_2(x)\}$. We use an alternative approach that allows us to work with just two bounding functions. In particular, we classify the space $\mathcal X$ according to the finite dimensional reduction map $\kappa : \mathcal X \to \mathbb R^3$ defined as follows:

$$\kappa(x) \stackrel{\text{def}}{=} \{q(x), \bar{q}(x), x(1)\}. \tag{5}$$

Relative to a SOPS's image under κ , we formally define bounding functions as follows.

Definition 1.6. Fix an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and a region $K \subset \mathbb{R}^3$. The functions $\ell_K, u_K : \mathbb{R} \to \mathbb{R}$ are bounding functions (associated with K) if

$$\ell_K(t) \le x(t) \le u_K(t)$$
, for all $t \in \mathbb{R}$,

whenever $x \in \mathcal{X}$ is a SOPS to Wright's equation at a parameter $\alpha \in I_{\alpha}$ satisfying $\kappa(x) \in K$.

In practice, we define the functions u_K , ℓ_K as piecewise constant functions, which are easy to represent and rigorously integrate on a computer. To ensure proper mathematical rigor and computational reliability, we have used interval arithmetic for the execution of our computer-assisted proofs [11, 17]. Notably, our algorithms use a rigorous numerical integrator for delay differential equations, about which there is a growing literature [1, 10, 20, 21]. These computational details are discussed further in Appendix A.

To summarize by Theorem 1.5, in order to prove that there is a unique SOPS, it is sufficient to show that every SOPS is asymptotically stable. This breaks into two major parts: characterizing SOPS to Wright's equation and bounding their Floquet multipliers. To accomplish the first part, we begin by constructing compact regions $K_1, K_2 \subset \mathbb{R}^3$, described in Algorithm 3.5 and Algorithm 3.8 respectively, for which $K_1 \cup K_2$ contains the κ -image of all SOPS to Wright's equation. We then use a branch and prune method, defined in Algorithm 5.1, to refine these initial global bounds. This algorithm branches by subdividing $K_1 \cup K_2$ into smaller pieces, and prunes by using Algorithm 2.2 to develop tighter bounding functions. The end result of this process is a collection $\mathcal A$ of subsets of $K \subset \mathbb{R}^3$, and in Theorem 5.2 we prove for a given parameter range $[\alpha_{min}, \alpha_{max}]$ that if $x \in \mathcal{X}$ is a SOPS then $\kappa(x) \in \bigcup_{K \in \mathcal{A}} K$. The task then becomes to show that every SOPS is asymptotically stable. For a given region $K \subset \mathbb{R}^3$, we use Algorithm 4.2 to derive a bound on the Floquet multipliers of any SOPS with κ -image contained in K. This is then combined with the branch and prune method in Algorithm 5.3. Finally, the proof to Theorem 1.3 is given in Section 6, where, in addition, we discuss the computational limitations of our approach.

2 A computational approach

Theorem 1.5 effectively transforms Jones's Conjecture (Conjecture 1.2) into the problem of studying the asymptotic dynamics of SOPS, and in turn, their Floquet multipliers / Lyapunov exponents. In a neighborhood about a periodic function, one can develop estimates on these Floquet multipliers [2, 26]. However these bounds rely significantly on this neighborhood about the periodic function being relatively small. In effect, Xie shows that any SOPS to Wright's equation is stable for each $\alpha \geq 5.67$ by first showing that all such solutions reside within a narrow region, and subsequently shows that all periodic orbits in that region are asymptotically stable. This first step is the more difficult part, and the reason Xie restricts his proof to $\alpha \geq 5.67$.

In Xie's thesis [26] a case-by-case analysis is used to obtain a region within which all SOPS must lie. Specifically, if $\bar{q}(x) \geq 3$ then asymptotic analysis [12] precisely describes the approximate form of the SOPS with tight error estimates. For the alternative case, Xie divided the possibility of $\bar{q}(x) < 3$ into several sub-cases and showed that each of these led to a contradiction when $\alpha \geq 5.67$. In our analysis we make similar assumptions by considering a SOPS's image under the map $\kappa(x) = \{q(x), \bar{q}(x), x(1)\}$ and the bounding functions associated with various regions $K \subset \mathbb{R}^3$. For any region $K \subset \mathbb{R}^3$ there is not a unique choice of bounding functions. In fact, we develop techniques which iteratively tighten the bounding functions for a fixed region K. If in our process

of tightening bounding functions we derive a contradiction, such as $\ell_K(t) > u_K(t)$, then we may conclude that there does not exist any SOPS x for which $\kappa(x) \in K$.

In performing a case-by-case analysis of SOPS to Wright's equation, we are principally concerned with bounding all possible SOPS, and we find it useful to introduce the notion of an I_{α} -exhaustive set.

Definition 2.1. Fix an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and consider a set $K \subset \mathbb{R}^3$. The set K is I_{α} -exhaustive if $\kappa(x) \in K$ for any SOPS $x \in \mathcal{X}$ to Wright's equation at parameter $\alpha \in I_{\alpha}$.

To derive a sufficiently small I_{α} -exhaustive set we employ techniques from global optimization theory. Specifically, we use a branch and prune algorithm which is derived from the classical global optimization technique of branch and bound [4,15,19]. Our branch and prune is designed so that it will output an I_{α} -exhaustive set, a result proved in Theorem 5.2.

The branch and prune algorithm begins with an initial finite set $S = \{K_i : K_i \subset \mathbb{R}^3\}$ for which $\bigcup_{K \in S} K$ is I_{α} -exhaustive. The construction of this initial set is described in Section 3, specifically in Algorithms 3.5 and 3.8. We then alternate between branching and pruning the elements of S. The branching subroutine divides an element $K \in S$ into two pieces K_A and K_B for which $K = K_A \cup K_B$, and then replaces K in the set S by the two smaller regions. The pruning algorithm uses a variety of techniques to derive sharper bounding functions on the region K. Furthermore, if we can prove that the preimage $\kappa^{-1}(K) \subseteq \mathcal{X}$ cannot contain any SOPS, then we remove the region K from the set S. The branch and prune algorithm terminates when the diameter of every region K is less than some preset constant.

In contrast to the prototypical optimization problem of bounding the minimum of an objective function, we are concerned with characterizing SOPS to Wright's equation. In particular, our pruning algorithm is designed to tighten the bounding functions associated with a region K, reduce the size of K, and to discard the region if we can prove that $\kappa^{-1}(K)$ does not contain any SOPS. The algorithm takes as input an interval I_{α} , a region K, and a pair of bounding functions u_K , ℓ_K . As output the algorithm produces a region $K' \subset K$ and a pair of bounding functions $u_{K'}$, $\ell_{K'}$. The set K is taken to be rectangular, that is $K = I_q \times I_{\bar{q}} \times I_M$ where $I_q = [q_{min}, q_{max}]$ and $I_{\bar{q}} = [\bar{q}_{min}, \bar{q}_{max}]$ and $I_M = [M_{min}, M_{max}]$. Additionally, this algorithm takes as input a computational parameter $n_{Time} \in \mathbb{N}$ relating to how we store the bounding functions u_K , ℓ_K on the computer (see Appendix A).

The six steps in the pruning algorithm (Algorithm 2.2) are independent of one another and can be implemented in any order. In Steps 1-4 we describe how to tighten the bounds on K, u_K and ℓ_K . Each step is constructed so that the output does not worsen the existing bounds. That is each step of the algorithm produces an output for which $K' \subseteq K$ and the inequalities $u_{K'} \leq u_K$ and $\ell_{K'} \geq \ell_K$ hold. At the end of each step we update our input so that we use the improved bounds in the next step. That is, we define:

$$K \stackrel{\text{def}}{=} K'$$
 $u_K \stackrel{\text{def}}{=} u_{K'}$ $\ell_K \stackrel{\text{def}}{=} \ell_{K'}$ (6)

and subsequently modify K', $u_{K'}$ and $\ell_{K'}$ as described in each individual step. In Steps 5-6, we check conditions which would imply that the region K cannot contain the κ -image of SOPS to Wright's equation. If this is the case, the algorithm returns $K = \emptyset$.

Algorithm 2.2 (Pruning Algorithm). This algorithm takes as input $I_{\alpha} = [\alpha_{min}, \alpha_{max}], K = [q_{min}, q_{max}] \times [\bar{q}_{min}, \bar{q}_{max}] \times [M_{min}, M_{max}] \subseteq \mathbb{R}^3$ and associated bounding functions ℓ_K and u_K , as well as the computational parameter $n_{Time} \in \mathbb{N}$. The outputs consist of a region $K' \subseteq \mathbb{R}^3$ and

associated bounding functions $\ell_{K'}$ and $u_{K'}$.

 $\begin{array}{l} \textit{Define } I_q = [q_{min}, q_{max}], \ I_{\bar{q}} = [\bar{q}_{min}, \bar{q}_{max}] \ \textit{and} \ I_M = [M_{min}, M_{max}] \ \textit{as well as } L_{min} \stackrel{\text{def}}{=} q_{min} + \bar{q}_{min}, \\ L_{max} \stackrel{\text{def}}{=} q_{max} + \bar{q}_{max} \ \textit{and} \ I_L \stackrel{\text{def}}{=} [L_{min}, L_{max}]. \end{array}$

1. We tighten the bounding functions associated with the region K using

$$u_{K'}(t) \stackrel{\text{def}}{=} \begin{cases} \min\{M_{max}, u_K(1)\} & \text{if } t = 1\\ \min\{0, u_K(t)\} & \text{if } t \in [-\bar{q}_{min}, 0] \cup [q_{max}, L_{min}] \\ u_K(t) & \text{otherwise} \end{cases}$$

$$\ell_{K'}(t) \stackrel{\text{def}}{=} \begin{cases} \max\{M_{min}, \ell_K(1)\} & \text{if } t = 1\\ \max\{0, \ell_K(t)\} & \text{if } t \in [-L_{min}, \bar{q}_{max}] \cup [0, q_{min}] \cup [L_{max}, L_{min} + q_{min}] \\ \ell_K(t) & \text{otherwise.} \end{cases}$$
(8)

Lastly we update our bounds using Line (6).

2. If x satisfies Wright's equation we can use variation of parameters to refine the bounding functions. For our computational parameter $n_{Time} \in \mathbb{N}$, we base our calculation about a collection of points separated by a uniform distance of $1/n_{Time}$. That is, define $\Delta = 1/n_{Time}$ and $I_{\Delta} \stackrel{\text{def}}{=} [0, \Delta]$, and fix $t_0 \in \{k \cdot \Delta\}_{k \in \mathbb{Z}}$ and $s \in I_{\Delta}$. We may refine the values of $u_K(t_0 + s)$, $u_K(t_0 - s)$, $\ell_K(t_0 + s)$, $\ell_K(t_0 - s)$ as follows:

$$u_{K''}(t_{0}+s) \stackrel{\text{def}}{=} u_{K}(t_{0}) + s \cdot \sup_{\alpha \in I_{\alpha}, r \in I_{\Delta}} \sup_{\ell_{K} \leq x \leq u_{K}} -\alpha \left(e^{x(t_{0}-1+r)} - 1 \right)$$

$$u_{K''}(t_{0}-s) \stackrel{\text{def}}{=} u_{K}(t_{0}) - s \cdot \inf_{\alpha \in I_{\alpha}, r \in I_{\Delta}} \inf_{\ell_{K} \leq x \leq u_{K}} -\alpha \left(e^{x(t_{0}-1-r)} - 1 \right)$$

$$\ell_{K''}(t_{0}+s) \stackrel{\text{def}}{=} \ell_{K}(t_{0}) + s \cdot \inf_{\alpha \in I_{\alpha}, r \in I_{\Delta}} \inf_{\ell_{K} \leq x \leq u_{K}} -\alpha \left(e^{x(t_{0}-1+r)} - 1 \right)$$

$$\ell_{K''}(t_{0}-s) \stackrel{\text{def}}{=} \ell_{K}(t_{0}) - s \cdot \sup_{\alpha \in I_{\alpha}, r \in I_{\Delta}} \sup_{\ell_{K} \leq x \leq u_{K}} -\alpha \left(e^{x(t_{0}-1-r)} - 1 \right)$$

and

$$u_{K'}(t_0 + s) \stackrel{\text{def}}{=} \min \{ u_K(t_0 + s), u_{K''}(t_0 + s) \}$$

$$u_{K'}(t_0 - s) \stackrel{\text{def}}{=} \min \{ u_K(t_0 - s), u_{K''}(t_0 - s) \}$$

$$\ell_{K'}(t_0 + s) \stackrel{\text{def}}{=} \max \{ \ell_K(t_0 + s), \ell_{K''}(t_0 + s) \}$$

$$\ell_{K'}(t_0 - s) \stackrel{\text{def}}{=} \max \{ \ell_K(t_0 - s), \ell_{K''}(t_0 - s) \}.$$

Appendix A explains in further detail the computational aspects of this step. Lastly we update our bounds using Line (6).

3. In this step we refine our bounds on I_q and I_M using u_K and ℓ_K . At t = q(x) the function x(t) changes sign from positive to negative. We sharpen the bounds on I_q by defining:

$$q'_{min} \stackrel{\text{def}}{=} \inf\{t \in I_q : \ell_K(t) \le 0\}$$

$$q'_{max} \stackrel{\text{def}}{=} \sup\{t \in I_q : u_K(t) \ge 0\}$$

$$I_{q'} \stackrel{\text{def}}{=} [q'_{min}, q'_{max}]. \tag{10}$$

Additionally we make the following refinement:

$$I_{M'} \stackrel{\text{def}}{=} [M_{min}, M_{max}] \cap [\ell_K(1), u_K(1)].$$

Lastly we define $K' \stackrel{\text{def}}{=} I_{q'} \times I_{\bar{q}} \times I_{M'}$ and update our bounds using Line (6).

4. If $x \in \mathcal{X}$ is a SOPS with period $L \in I_L$, then x(t) = x(t + L). Using this relation, we make the following refinement:

$$\ell_{K'}(t) \stackrel{\text{def}}{=} \max \left\{ \ell_K(t), \min_{L' \in I_L} \ell_K(t + L') \right\}$$
(11)

$$u_{K'}(t) \stackrel{\text{def}}{=} \min \left\{ u_K(t), \max_{L' \in I_L} u_K(t + L') \right\}. \tag{12}$$

Lastly we update our bounds using Line (6) as appropriate.

- 5. If there is some point $t \in \mathbb{R}$ for which $\ell_K(t) > u_K(t)$ then RETURN $K' \stackrel{\text{def}}{=} \emptyset$.
- 6. If $\min_{t \in I_q} \ell_K(t+1) > -\log \frac{\alpha_{min}}{\pi/2}$, then RETURN $K' \stackrel{\text{def}}{=} \emptyset$.

Proposition 2.3. Let $I_{\alpha} = [\alpha_{min}, \alpha_{max}], \ \alpha_{min} \geq \frac{\pi}{2},$

$$K = [q_{min}, q_{max}] \times [\bar{q}_{min}, \bar{q}_{max}] \times [M_{min}, M_{max}] \subseteq \mathbb{R}^3$$

and u_K, ℓ_K be input for Algorithm 2.2 with any computational parameter $n_{Time} \in \mathbb{N}$. Suppose that $x \in \mathcal{X}$ is a SOPS at parameter $\alpha \in I_{\alpha}$, and let $\{K', u_{K'}, \ell_{K'}\}$ be the result of Algorithm 2.2. If $\kappa(x) \in K$, then $\kappa(x) \in K'$ and $\ell_{K'} \leq x \leq u_{K'}$.

Proof. We prove that Proposition 2.3 holds for each step of the algorithm individually. Given an interval $I_{\alpha} \subset \mathbb{R}$, let $x \in \mathcal{X}$ be a SOPS at parameter $\alpha \in I_{\alpha}$.

- 1. Recall $\kappa(x) = \{q(x), \bar{q}(x), \max(x)\}$. Since $\max_{t \in \mathbb{R}} x = x(1)$ then the refinements in (7) and (8) for the case in which t = 1 are appropriate. For the other two refinements in each equation note that by definition a function $x \in \mathcal{X}$ is non-negative on the interval [0, q(x)] and non-positive on the interval $[q(x), q(x) + \bar{q}(x)]$. Hence, x is non-negative on $[0, q_{min}]$ and non-positive on $[q_{min}, L_{min}]$. If x is a SOPS then it has period $L = q(x) + \bar{q}(x)$ and we may further conclude that it is non-negative on the intervals $[-L, \bar{q}(x)]$ and [L, L + q(x)], and non-positive on the interval $[-\bar{q}(x), 0]$. Hence, x is non-negative on $[-L_{min}, \bar{q}_{max}]$ and $[L_{max}, L_{min} + q_{min}]$, and non-positive on $[-\bar{q}_{min}, 0]$. The refinements in (7) and (8) reflect these restrictions.
- 2. To estimate an upper bound on $x(t_0 + s)$, we apply variation of parameters to Wright's equation, obtaining

$$x(t_0 + s) = x(t_0) + \int_{t_0}^{t_0 + s} -\alpha \left(e^{x(r-1)} - 1 \right) dr.$$
 (13)

Taking the Riemann upper sum of this integral with step size s, we deduce that $x(t_0 + s)$ is bounded above by the RHS of (9). As $x(t_0 + s) \le u_K(t_0 + s)$ it follows that $x(t_0 + s) \le \min\{u_K(t_0 + s), u_{K''}(t_0 + s)\}$. The proofs for the refinements of $u_K(t_0 - s), \ell_K(t_0 + s), \ell_K(t_0 - s)$ follow with parity.

3. Let $x \in \mathcal{X}$ be such that $\kappa(x) \in K$. From our definitions of q'_{min} and q'_{max} it follows that

$$x(t) \ge \ell_K(t) > 0$$
, for all $t \in (q_{min}, q'_{min})$,
 $x(t) \le u_K(t) < 0$, for all $t \in (q'_{max}, q_{max})$.

Hence it follows that $x(t) \neq 0$ for $t \in (q_{min}, q'_{min}) \cup (q'_{max}, q_{max})$. Since $q(x) \in I_q$, it must follow that $q(x) \in [q'_{min}, q'_{max}]$, thus justifying the refinement in (10). Regarding the refinement of I_M , as $\ell_K(1) \leq x(1) \leq u_K(1)$ it clearly follows that $[\kappa(x)]_3 = x(1) \in I_{M'}$.

4. If x is periodic with period L, then x(t) = x(t + L). Since $L \in I_L$ then we may derive upper/lower bounds on x(t + L) as follows:

$$\min_{L'\in I_L}\ell_K(t+L')\leq \min_{L'\in I_L}x(t+L')\leq x(t+L)\leq \max_{L'\in I_L}x(t+L')\leq \max_{L'\in I_L}u_K(t+L').$$

Hence it follows that $\min_{L' \in I_L} \ell_K(t + L') \leq \ell_K(t)$ and $u_K(t) \leq \max_{L' \in I_L} u_K(t + L')$, thus justifying our refinements in (11) and (12).

- 5. If $\ell_K(t) > u_K(t)$, then it is impossible for any $x \in \mathcal{X}$ to satisfy $\ell_K(t) \leq x(t) \leq u_K(t)$. Since u_K , and ℓ_K are bounding functions associated with K, this contradiction leads us to conclude that there cannot exist any SOPS $x \in \mathcal{X}$ for which $\kappa(x) \in K$.
- 6. By the results in [24], if x is a SOPS to Wright's equation and $\alpha \geq \frac{\pi}{2}$, then

$$\min x \le -\log \frac{\alpha}{\pi/2}.\tag{14}$$

If $x \in \mathcal{X}$ is a SOPS then $\min_{t \in \mathbb{R}} x(t) = x(q+1)$, whereby $\min_{t \in \mathbb{R}} x(t) > \min_{t \in I_q} \ell_k(t+1)$. Hence, if $\min_{t \in I_q} \ell_k(t+1) > -\log \frac{\alpha_{min}}{\pi/2}$ then (14) is violated, and so there cannot exist any SOPS $x \in \mathcal{X}$ for which $\kappa(x) \in K$.

3 Initial Bounds on SOPS to Wright's Equation

In order to apply the branch and prune algorithm, we must first construct an initial I_{α} -exhaustive set. Due to the sustained interest in Wright's equation, there are considerable a priori estimates we can employ to describe slowly oscillating solutions [6,12,25]. Since considerably sharper estimates are obtained under the assumption $\bar{q} \geq 3$, we will construct two regions K_1 and K_2 corresponding to SOPS $x \in \mathcal{X}$ for which $\bar{q}(x) \leq 3$ and $\bar{q}(x) \geq 3$ respectively. Taken together $K_1 \cup K_2$ will form an I_{α} -exhaustive set, which we prove in Corollary 3.12.

While sharper estimates are available for additional sub-cases [12], we present a collection of these estimates we have found sufficient for our purposes. Note that these lemmas are not a verbatim reproduction. We have translated results applicable to the quadratic form of Wright's equation given in (1) so that they apply to the exponential form of Wright's equation given in (2).

Lemma 3.1 (See [25]). Let $x \in \mathcal{X}$ be a solution to Wright's equation at parameter $\alpha > 0$. Then

$$-\alpha(e^{\alpha}-1) < x(t) < \alpha$$

for all t > 0.

Lemma 3.2 (See [6, Theorem 3.1]). Let $\alpha > e^{-1}$ and suppose that $x \in \mathcal{X}$ and is a solution to Wright's equation. We construct a sequence of functions $p_i : (-\infty, 1] \to \mathbb{R}$ for $i = 1, 2, \cdots$ by setting $p_1(t) = \alpha t$ and recursively defining:

$$p_{i+1}(t) \stackrel{\text{def}}{=} -\alpha \int_0^t \left(e^{p_i(s-1)} - 1 \right) ds.$$

For example $p_2(t) = \alpha t + e^{-\alpha} - e^{\alpha(t-1)}$. Then $x(t) > p_i(t)$ for t < 0, and $x(t) < p_i(t)$ for $t \in (0,1]$. Furthermore $x(t) < p_i(1)$ for all $t \ge 0$. Additionally $|p_i(t)|$ is increasing in α .

Lemma 3.3 (See [6, Theorem 3.4]). Let $x \in \mathcal{X}$ and suppose that $\bar{q} \geq 3$ and that $\alpha \geq \frac{\pi}{2}$. Define $a_1(\alpha) = -(\alpha - 1)$ and the recursive relation $a_{i+1}(\alpha) = \alpha(e^{a_i(\alpha)} - 1)$. Then $x(t) < -t \cdot a_i(\alpha)$ for $t \in [-1,0)$ and $i \in \mathbb{N}$.

Lemma 3.4. Suppose that $x \in \mathcal{X}$ is a SOPS to Wright's Equation. If $\alpha \geq \frac{\pi}{2}$ then

$$1 + \frac{1}{\alpha} \left(\frac{\alpha + e^{-\alpha} - 1}{\exp\{\alpha + e^{-\alpha} - 1\} - 1} \right) < q < 2 + \frac{1}{\alpha}$$

$$1 + \frac{1}{\alpha} < \bar{q} < \max\{3, 2 + \left| \frac{e^{\alpha} - 1}{e^{a_i(\alpha)} - 1} \right| \}$$

where $a_i(\alpha)$ is taken as in Lemma 3.3. Additionally, if $\alpha \geq 2$ then q < 2.

Proof. All but the upper bound on q follows from Theorem 3.5 in [6]. To prove the upper bound, assume that $q \geq 2$, and consider the quadratic version of Wright's equation given in (1). It follows that $y(t) \geq y(2)$ for all $t \in [1,2]$, and thereby $y'(t) \geq -\alpha y(2)[1+0]$ for all $t \in [2,3]$. From this we obtain $q < 2 + \frac{1}{\alpha}$.

Step 2 constructs iterative bounds analogous to Lemmas 3.2 and 3.3. When $\bar{q}(x) \geq 3$, Step 3 obtains bounds on I_M and I_q which are tighter than the bounds given in Lemma 3.1 and Lemma 3.4. Additionally, the branching procedure further reduces the size of I_q , $I_{\bar{q}}$, and I_M . Below in Algorithm 3.5 we construct the initial bounds for a region $K \subseteq \mathbb{R}^3$ containing the κ -image of SOPS $x \in \mathcal{X}$ for which $\bar{q}(x) \leq 3$.

Algorithm 3.5. The input we take is an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and computational parameters $i_0, n_{Time} \in \mathbb{N}$. The output is a rectangle $K = I_q \times I_{\bar{q}} \times I_M \subseteq \mathbb{R}^3$ and bounding functions u_K , ℓ_K .

1. Make the following definitions:

$$q_{min} \stackrel{\text{def}}{=} 1 + \inf_{\alpha \in I_{\alpha}} \frac{1}{\alpha} \left(\frac{\alpha + e^{-\alpha} - 1}{\exp{\{\alpha + e^{-\alpha} - 1\}} - 1} \right)$$

$$M_{min} \stackrel{\text{def}}{=} \inf_{\alpha \in I_{\alpha}} \log \left(1 + \alpha^{-1} \log \frac{\alpha}{\pi/2} \right)$$

$$I_{q} \stackrel{\text{def}}{=} \begin{cases} [q_{min}, 2] & \text{if } \alpha_{min} \ge 2\\ [q_{min}, 2 + \frac{1}{\alpha_{min}}] & \text{otherwise} \end{cases}$$

$$I_{\bar{q}} \stackrel{\text{def}}{=} \left[1 + \frac{1}{\alpha_{max}}, 3 \right]$$

$$I_{M} \stackrel{\text{def}}{=} [M_{min}, p_{i_{0}}(1)].$$

2. For p_i given as in Propositions 3.2, define bounding functions

$$\ell_K(t) \stackrel{\text{def}}{=} \begin{cases} 0 & \text{if } t = 0 \\ -\alpha_{max}(e^{\alpha_{max}} - 1) & \text{otherwise} \end{cases} \qquad u_K(t) \stackrel{\text{def}}{=} \begin{cases} 0 & \text{if } t = 0 \\ p_{i_0}(1) & \text{otherwise}. \end{cases}$$

These bounding functions are stored on the computer with time resolution n_{Time} as described in Appendix A.

Proposition 3.6. Fix an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ such that $\alpha_{min} \geq \frac{\pi}{2}$. Let K, u_K, ℓ_K denote the output of Algorithm 3.5. If $x \in \mathcal{X}$ is a SOPS to Wright's equation at parameter $\alpha \in I_{\alpha}$ and $\bar{q}(x) \leq 3$ then $\kappa(x) \in K$ and $\ell_K \leq x \leq u_K$.

Proof. We treat the two steps in order.

- 1. If $x \in \mathcal{X}$ is a SOPS to Wright's equation and $\bar{q}(x) \leq 3$, then by Lemma 3.4 it follows that $q(x) \in I_q$ and $\bar{q}(x) \in I_{\bar{q}}$. By Proposition 3.2 it follows that $x(1) \leq p_{i_0}(1)$. If x is a SOPS to Wright's equation with $\alpha \geq \pi/2$ then $\min x \leq -\log \frac{2\alpha}{\pi}$ [24]. If $\max x < \log \left(1 + \alpha^{-1} \log \frac{\alpha}{\pi/2}\right)$ then by integrating Wright's equation forward from t = q to t = q+1 it follows that $x(q+1) = \min x \leq -\log \frac{\alpha}{\pi/2}$, a contradiction. Hence we may assume that $x(1) \geq \log \left(1 + \alpha^{-1} \log \frac{\alpha}{\pi/2}\right) \geq M_{min}$.
- 2. Since $x \in \mathcal{X}$ then x(0) = 0, and by Lemma 3.1 and Proposition 3.2 it follows that $-\alpha(e^{\alpha} 1) \le x \le p_{i_0}(1)$ for any SOPS $x \in \mathcal{X}$. Hence ℓ_K and u_K are bounding functions for $K = I_q \times I_{\bar{q}} \times I_M$.

To construct the initial bounds for the case $\bar{q}(x) \geq 3$, we make greater use of a priori bounds. Unfortunately the bounds on $I_{\bar{q}}$ given in Lemma 3.4 are not sharp, that is the width of this estimate of $I_{\bar{q}}$ is greater than $e^{\alpha} - 2$. Using this estimate would be computational difficult. In [12] Nussbaum estimates the value of \bar{q} up to $\mathcal{O}(\frac{1}{\alpha})$ in the case of $\bar{q}(x) \geq 3$ and $\alpha \geq 3.8$. We derive a similar estimate which only assumes $\bar{q}(x) \geq 2$ and $\alpha > 0$. This estimate is better suited for numerical applications, and only needs bounds $\ell(t) \leq x(t) \leq u(t)$ that are defined over the time domain $t \in [-1, 4]$.

Lemma 3.7. Fix some $\alpha > 0$ and suppose that $x \in \mathcal{X}$ is a SOPS to Wright's equation, and let $\ell, u : \mathbb{R} \to \mathbb{R}$ be functions for which $\ell(t) \leq x(t) \leq u(t)$. Let $I_q \subset \mathbb{R}$ be an interval for which $q(x) \in I_q$ and suppose that $\bar{q}(x) \geq 2$. Define the following integral bounds:

$$U^{+} \stackrel{\text{def}}{=} \sup_{q \in I_{q}} \int_{q-1}^{q} \max \left\{ e^{u(t)} - 1, 0 \right\} dt \qquad U_{1}^{-} \stackrel{\text{def}}{=} \sup_{q \in I_{q}} \int_{q}^{q+1} - \min \left\{ e^{\ell(t)} - 1, 0 \right\} dt \qquad (15)$$

$$L^{+} \stackrel{\text{def}}{=} \inf_{q \in I_{q}} \int_{q-1}^{q} \max \left\{ e^{\ell(t)} - 1, 0 \right\} dt \qquad L_{1}^{-} \stackrel{\text{def}}{=} \inf_{q \in I_{q}} \int_{q}^{q+1} - \min \left\{ e^{u(t)} - 1, 0 \right\} dt \qquad (16)$$

and define $m \stackrel{\text{def}}{=} \min_{t \in I_q} \ell(t+1)$. Then \bar{q} is bounded by the inequalities

$$2 + \frac{L^{+} - U_{1}^{-}}{|e^{m} - 1|} \le \bar{q} \le 2 + \frac{U^{+} - L_{1}^{-}}{|e^{u(-1)} - 1|}.$$
(17)

The proof is delayed until the end of this section. The computational details of how we evaluate the integrals in (15) and (16) are discussed in Appendix A. Below in Algorithm 3.8 we construct the initial bounds for a region $K \subseteq \mathbb{R}^3$ containing the κ -image of SOPS $x \in \mathcal{X}$ for which $\bar{q}(x) \geq 3$.

Algorithm 3.8. The input is an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and computational parameters $i_0, j_0, n_{Time}, N_{period} \in \mathbb{N}$. The output is a rectangle $K = I_q \times I_{\bar{q}} \times I_M$ and bounding functions u_K, ℓ_K .

1. Make the following definitions for $K = I_q \times I_{\bar{q}} \times I_M$:

$$q_{min} \stackrel{\text{def}}{=} 1 + \inf_{\alpha \in I_{\alpha}} \frac{1}{\alpha} \left(\frac{\alpha + e^{-\alpha} - 1}{\exp\{\alpha + e^{-\alpha} - 1\} - 1} \right)$$

$$I_{q} \stackrel{\text{def}}{=} \left[q_{min}, 2 + \frac{1}{\alpha_{min}} \right]$$

$$I_{\bar{q}} \stackrel{\text{def}}{=} \left[3, \sup_{\alpha \in I_{\alpha}} 2 + \left| \frac{e^{\alpha} - 1}{e^{a_{j_{0}}(\alpha)} - 1} \right| \right]$$

$$I_{M} \stackrel{\text{def}}{=} \left[0, p_{i_{\alpha}}(1) \right]$$

where $a_i(\alpha)$ is taken as in Lemma 3.3.

2. For p_i and a_j given as in Propositions 3.2 and 3.3 respectively, define bounding functions ℓ_K and u_K

$$\ell_K(t) \stackrel{\text{\tiny def}}{=} \begin{cases} 0 & \text{if } t = 0 \\ p_{i_0}(t) & \text{if } t < 0 \\ \inf_{\alpha \in I_\alpha} -\alpha(e^\alpha - 1) & \text{otherwise} \end{cases} \quad u_K(t) \stackrel{\text{\tiny def}}{=} \begin{cases} 0 & \text{if } t = 0 \\ \sup_{\alpha \in I_\alpha} -t \cdot a_{j_0}(\alpha) & \text{if } t \in [-1,0) \\ p_{i_0}(1) & \text{otherwise}. \end{cases}$$

These bounding functions are stored on the computer with time resolution n_{Time} as described in Appendix A.

- 3. Refine u_K and ℓ_K according to Step 1 of Algorithm 2.2. For N_{period} iterations, refine u_K and ℓ_K according to Step 2 of Algorithm 2.2 for values $t_0 \in [-4, 4]$. Then define I_q and I_M according to Step 3 of Algorithm 2.2.
- 4. For values of $m, L^+, L_1^-, U^+, U_1^-$ given as in Proposition 3.7, define:

$$\bar{q}_{min} \stackrel{\text{def}}{=} 2 + \frac{L^+ - U_1^-}{|e^m - 1|}, \qquad \bar{q}_{max} \stackrel{\text{def}}{=} 2 + \frac{U^+ - L_1^-}{|e^{u_K(-1)} - 1|}.$$

If $\bar{q}_{max} < 3$ then define $K = \emptyset$. Otherwise define $I_{\bar{q}} = [\bar{q}_{min}, \bar{q}_{max}]$ and $K = I_q \times I_{\bar{q}} \times I_M$.

Remark 3.9. In practice we select $i_0 = 2$ and $j_0 = 20$ in Step 2, which have proved sufficient for our purposes. In [6] the expressions for p_i are given in closed form for i = 1, 2, 3, 4, each function being increasingly complex. The sequence $a_j(\alpha)$ is convergent, and we use $j_0 = 20$ because we have found negligible improvements when using a larger index.

Proposition 3.10. Fix an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ such that $\alpha_{min} \geq \frac{\pi}{2}$, and fix computational parameters $i_0, j_0, n_{Time}, N_{period} \in \mathbb{N}$. Let $\{K, u_K, \ell_K\}$ denote the output of Algorithm 3.8. If $x \in \mathcal{X}$ is a SOPS to Wright's equation and $\bar{q}(x) \geq 3$ then $\kappa(x) \in K$ and $\ell_K \leq x \leq u_K$.

Proof. Let x be as described above. We describe the effect of each step of the algorithm in turn.

- 1. For $I_q, I_{\bar{q}}$ and I_M defined in Step 1, it follows from Lemma 3.4 that $q(x) \in I_q$ and $\bar{q}(x) \in I_{\bar{q}}$, and it follows from Lemma 3.1 and Lemma 3.2 that $x(1) \in I_M$.
- 2. Since $x \in \mathcal{X}$ then x(0) = 0. By Lemma 3.1 then any SOPS $x \in \mathcal{X}$ satisfies the inequality $-\alpha(e^{\alpha} 1) \leq x(t) \leq p_{i_0}(1)$. The definition of the ℓ_K bound for t < 0 follows from Lemma 3.2, and the definition of the u_K bound for $t \in [-1,0)$ follows from Lemma 3.3.
- 3. The results of Steps 1 and 2 produce a region K with bounding functions u_K , ℓ_K for which $\kappa(x) \in K$ whenever there is a SOPS $x \in \mathcal{X}$ satisfying $\bar{q}(x) \geq 3$. By Proposition 2.3, implementing Steps 2 and 3 of Algorithm 2.2 preserves this property.
- 4. Since $\bar{q}(x) \geq 3 > 2$ then by Lemma 3.7 it follows that $\bar{q}_{min} \leq \bar{q}(x) \leq \bar{q}_{max}$. If $\bar{q}_{max} < 3$, this contradicts our initial assumption that $\bar{q}(x) \geq 3$, whereby there are no SOPS $x \in \mathcal{X}$ to Wright's equation at any parameter $\alpha \in I_{\alpha}$ for which $\bar{q}(x) \geq 3$. Otherwise for our definition of $K = I_q \times I_{\bar{q}} \times I_M$ it follows that $\kappa(x) \in K$ whenever $\bar{q}(x) \geq 3$.

We present an application of this theorem.

Proposition 3.11. If $x \in \mathcal{X}$ is a SOPS to (1) and $\alpha \in \left[\frac{\pi}{2}, 2.07\right]$ then $\bar{q}(x) < 3$.

Proof. First we constructed subintervals I_{α} of [1.57, 2.07] of width 0.1, and for each subinterval I_{α} we ran Algorithm 3.8 with computational parameters $i_0 = 2$, $j_0 = 20$, $n_{time} = 128$, and $N_{period} = 10$ (see [28] for associated MATLAB code). In each case the algorithm returned $K = \emptyset$.

Corollary 3.12. Fix an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ such that $\alpha_{min} > \frac{\pi}{2}$, and fix computational parameters $i_0, j_0, N_{period} \in \mathbb{N}$. Let $\{K_1, u_{K_1}, \ell_{K_1}\}$ denote the output of Algorithm 3.5 and let $\{K_2, u_{K_2}, \ell_{K_2}\}$ the output of Algorithm 3.8. Then $K_1 \cup K_2$ is I_{α} -exhaustive.

Proof. Suppose that $x \in \mathcal{X}$ is SOPS to Wrights equation. If $\bar{q}(x) \leq 3$, then by Proposition 3.6 it follows that $\kappa(x) \in K_1$. If $\bar{q}(x) \geq 3$, then by Proposition 3.10 it follows that $\kappa(x) \in K_2$. Hence the set $K_1 \cup K_2$ is I_{α} -exhaustive.

Proof of Lemma 3.7. Let p denote the period of a SOPS $x \in \mathcal{X}$. By assumption x(p) = x(q) = 0, so by the fundamental theorem of calculus we have that for any SOPS x,

$$0 = x(p) - x(q) = \int_{q}^{p} x'(t)dt = \int_{q}^{p} -\alpha(e^{x(t-1)} - 1)dt = \int_{q-1}^{p-1} (e^{x(t)} - 1)dt.$$

Recall that any SOPS x(t) is positive for $t \in (0, q)$ and negative for $t \in (q, p)$. Hence the integrand above is positive on (q - 1, q) and negative on (q, p - 1), thus producing the following estimate:

$$\int_{q-1}^{q} |e^{x(t)} - 1| dt = \int_{q}^{p-1} |e^{x(t)} - 1| dt.$$
 (18)

For $t \in (q-1,q)$ the function x(t) is positive, whereby $|e^{x(t)}-1| = \max\{e^{x(t)}-1,0\}$. For the definitions of L^+ and U^+ given in (15) and (16), it follows that L^+ and U^+ bound the LHS of (18) as described below:

$$L^{+} \leq \int_{q-1}^{q} \max\{e^{\ell(t)} - 1, 0\}dt \leq \int_{q-1}^{q} |e^{x(t)} - 1|dt \leq \int_{q-1}^{q} \max\{e^{u(t)} - 1, 0\}dt \leq U^{+}.$$

We estimate the RHS of (18) using the two sums below:

$$L_1^- + L_2^- \le \int_q^{p-1} |e^{x(t)} - 1| dt \le U_1^- + U_2^-$$

where the constants $L_1^-, L_2^-, U_1^-, U_2^-$ are appropriately defined so that

$$L_1^- \le \int_q^{q+1} |e^{x(t)} - 1| dt \le U_1^- \tag{19}$$

$$L_2^- \le \int_{q+1}^{p-1} |e^{x(t)} - 1| dt \le U_2^-.$$
 (20)

For $t \in (q, q+1)$ the function x(t) is negative, whereby $|e^{x(t)} - 1| = -\min\{e^{x(t)} - 1, 0\}$. It follows from the definitions of L_1^- and U_1^- given in (15) and (16) that (19) is satisfied. To define L_2^- and U_2^- note that for the time period $t \in [q+1, p-1]$ we have that x'(t) > 0, whereby

$$x(t) \ge x(q+1)$$

 $x(t) \le x(p-1) = x(-1) \le u(-1).$

By definition $m \le x(q+1)$, and as $p-q=\bar{q}$ we can then define

$$U_2^- \stackrel{\text{def}}{=} \int_{q+1}^{p-1} |e^m - 1| dt \qquad \qquad L_2^- \stackrel{\text{def}}{=} \int_{q+1}^{p-1} |e^{u(-1)} - 1| dt$$
$$= (\bar{q} - 2)|e^m - 1| \qquad \qquad = (\bar{q} - 2)|e^{u(-1)} - 1|.$$

Using these definitions, (20) is satisfied. From (18), we get the following upper and lower bounds on \bar{q} , from which (17) follows.

$$\begin{split} L_1^- + L_2^- &\leq U^+ & U_1^- + U_2^- \geq L^+ \\ (\bar{q} - 2)|e^{u(-1)} - 1| &\leq U^+ - L_1^- & (\bar{q} - 2)|e^m - 1| \geq L^+ - U_1^- \\ &\bar{q} \leq 2 + \frac{U^+ - L_1^-}{|e^{u(-1)} - 1|} & \bar{q} \geq 2 + \frac{L^+ - U_1^-}{|e^m - 1|}. \end{split}$$

4 Bounding the Floquet Multipliers.

In this section we describe how to estimate the Floquet multipliers of SOPS contained within the bounds derived in Sections 2 and 3. This method follows the approach of [26] with modifications to take advantage of numerical computations. To review this method, we first define a hyperplane in C[-1,0] as

$$H \stackrel{\text{\tiny def}}{=} \{\varphi \in C[-1,0] : \varphi(0) = 0\}.$$

For a function y we define $y_0 \in C[-1,0]$ to be the cut-off function of y on [-1,0], and for a constant $L \in \mathbb{R}$ we define $y_L \stackrel{\text{def}}{=} [y(t+L)]_0$. Locally, one can construct a smooth Poincaré map $\Phi: H \to H$ via the solution operator. If $x \in \mathcal{X}$ is a SOPS, then x_0 is a fixed point of Φ , and the Floquet multipliers of x are the eigenvalues of $D_{\varphi}\Phi(x_0)$. Of course x_0 is a trivial eigenfunction with associated eigenvalue $\lambda = 1$. The nontrivial and nonzero Floquet multipliers of the SOPS can be calculated by solving the following boundary value problem:

Theorem 4.1 (See [26, Theorem 2.2.3]). Suppose that $x \in \mathcal{X}$ is a SOPS to (2) with period L. Define the linearized DDE below:

$$y'(t) = -\alpha e^{x(t-1)}y(t-1). (21)$$

Then $\lambda \neq 0$ is a nontrivial eigenvalue of $D_{\varphi}\Phi(x_0)$ if and only if (22) has a nonzero solution $h \in H$ for which

$$-y(L)\frac{x'_{L}(t)}{x'(L)} + y_{L}(t) = \lambda h(t)$$
(22)

where the function h is then an eigenfunction of $D_{\varphi}\Phi(x_0)$ associated with λ , and y(t) solves (21) with initial condition $y_0 = h$.

We are able to bound the Floquet multipliers by studying this boundary value problem defined in (21) and (22), a calculation which is systematized through Algorithm 4.2. If this algorithm outputs a value $\Lambda_{max} < 1$ then all SOPS $x \in \kappa^{-1}(K)$ are asymptotically stable. We are able to improve upon Xie's method in [26, 27] by repeating certain steps, somewhat analogous to the recursive bounds defined in Lemma 3.2. The great advantage for doing this numerically as opposed to analytically is that these repetitions while tedious and time consuming for the mathematician are "effortless" for the computer.

Algorithm 4.2. Fix $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and $K = [q_{min}, q_{max}] \times [\bar{q}_{min}, \bar{q}_{max}] \times [M_{min}, M_{max}] \subseteq \mathbb{R}^3$ with associated bounding functions u_K, ℓ_K . Furthermore, fix computational parameters n_{Time} , $N_{Floquet}$, $M_{Floquet} \in \mathbb{N}$. The output of the algorithm is $\Lambda_{max} \in \mathbb{R}_+$.

1. Define $L_{min} \stackrel{\text{def}}{=} q_{min} + \bar{q}_{min}$, $L_{max} \stackrel{\text{def}}{=} q_{max} + \bar{q}_{max}$ and $I_L \stackrel{\text{def}}{=} [L_{min}, L_{max}]$, and define the function $Y : [-1, 0] \to \mathbb{R}$ by

$$Y(t) \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } t \in [-1, 0) \\ 0 & \text{if } t = 0. \end{cases}$$

2. Extend the function $Y: [-1, L_{max}] \to \mathbb{R}$ by

$$Y(t) \stackrel{\text{def}}{=} \alpha_{max} \int_0^t \left(Y(s-1) \sup_{\ell_K \le x \le u_K} e^{x(s-1)} \right) ds \qquad if \ t \ge 0, \tag{23}$$

evaluating the integral using an upper Riemann sum with a uniform step size of $1/n_{Time}$. Appendix A discusses in further details how we compute this integral.

3. Define Z as below:

$$Z(t) \stackrel{\text{def}}{=} \left(\max_{L \in [L_{min}, L_{max}]} Y(L) \right) \max_{\ell_K \le x \le u_K} \left| \frac{e^{x(t-1)} - 1}{e^{x(-1)} - 1} \right| + Y(t).$$

4. For $t \in [-L_{min}, 0]$ define Z_L as below:

$$Z_L(t) \stackrel{\text{def}}{=} \max_{L_{min} \le L \le L_{max}} Z(t+L).$$

5. For $t \in [-(L_{min} - 1), 0]$ refine the function Z_L by

$$Z'_{L}(-t) \stackrel{\text{def}}{=} \alpha_{max} \int_{-t}^{0} \left(Z_{L}(s-1) \sup_{\ell_{K} \leq x \leq u_{K}} e^{x(s-1)} \right) ds$$

$$Z_{L}(-t) \stackrel{\text{def}}{=} \min \left\{ Z_{L}(-t), Z'_{L}(-t) \right\},$$

$$(24)$$

evaluating the integral using an upper Riemann sum with a uniform step size of $1/n_{Time}$. Appendix A discusses in further details how we compute this integral.

Repeat this step $M_{Floquet}$ number of times.

6. Define

$$\Lambda_{max} \stackrel{\text{def}}{=} \sup_{t \in [-1,0]} Z_L(t).$$

- 7. If $\Lambda_{max} < 1$ then STOP.
- 8. Otherwise define

$$Y(t) \stackrel{\text{def}}{=} \min\left\{1, Z_L(t)\right\}, \qquad \text{for } t \in [-1, 0]$$
 (25)

and GOTO Step 2. After reaching this step $N_{Floquet}$ times, exit the program.

Theorem 4.3. Fix $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and $K = [q_{min}, q_{max}] \times [\bar{q}_{min}, \bar{q}_{max}] \times [M_{min}, M_{max}] \subseteq \mathbb{R}^3$. If Algorithm 4.2 terminates with $\Lambda_{max} < 1$, then all SOPS $x \in \mathcal{X}$ satisfying $\kappa(x) \in K$ must be asymptotically stable. If the algorithm terminates having never reached Step 8, then the norm of all nontrivial Floquet multiplier are bounded above by Λ_{max} .

Proof. Fix some $x \in \mathcal{X}$ for which $\kappa(x) \in K$. By the definition made in Step 1, the period of x is some $L \in I_L$. We use Theorem 4.1 to estimate the range of Floquet multipliers of x. That is, fix $\lambda \in \mathbb{C}$ and $h \in H$ and suppose that (λ, h) is a solution to (22). Define y(t) to be the solution of (21) through h, define z as

$$z(t) \stackrel{\text{def}}{=} -y(L)\frac{x'(t)}{x'(L)} + y(t) \tag{26}$$

and define $z_L(t) \stackrel{\text{def}}{=} z(t+L)$. Hence (λ, h) is a solution to (22) if and only if $z_L(t) = \lambda h(t)$ for $t \in [-1,0]$. As (21) is a linear DDE, we may assume without loss of generality that $\sup_{t \in [-1,0]} |h(t)| = 1$. Thereby, it follows that

$$|\lambda| = \sup_{t \in [-1,0]} |z_L(t)|.$$
 (27)

If we can show that the RHS of (27) is less than 1 uniformly for $x \in \kappa^{-1}(K)$, then we will have proven that all such SOPS are asymptotically stable. We prove that Steps 1-7 of Algorithm 4.2 produce functions Y, Z and Z_L and a bound Λ_{max} which satisfy the following inequalities uniformly for $x \in \kappa^{-1}(K)$

$$|y(t)| \le Y(t),$$
 $|z(t)| \le Z(t),$ $|z_L(t)| \le Z_L(t),$ $|\lambda| \le \Lambda_{max}.$

We describe the results of each step of Algorithm 4.2 in order, and then discuss how Step 8 affects what we may deduce about the output Λ_{max} .

1. By definition, if $h \in H$ then h(0) = 0, and by assumption $|h(t)| \le 1$ for $t \in [-1, 0]$. Thereby our definition of Y(t) in Step 1 satisfies $|y(t)| \le Y(t)$ for $t \in [-1, 0]$.

2. By definition y solves the linear DDE in (21). By variation of parameters it follows that

$$y(t) = \int_0^t -\alpha e^{x(s-1)}y(s-1)ds$$

for all $t \geq 0$. Equation (23) follows from this by taking a supremum over $\alpha \in I_{\alpha}$ and $\ell_K \leq x \leq u_K$. Thereby, Step 2 produces a function Y satisfying $|y(t)| \leq Y(t)$ for $t \geq 0$.

3. Step 3 defines a function Z to bound the norm of z defined in (26). As $x'(t) = -\alpha(e^{x(t-1)} - 1)$ it follows that

$$|z(t)| \le \left| y(L) \frac{e^{x(t-1)} - 1}{e^{x(L-1)} - 1} \right| + |y(t)|.$$

By periodicity, we may replace x'(L-1) with x'(-1). By taking a supremum over $L \in I_L$ and $\ell_K \leq x \leq u_K$, it follows that the function defined in Step 3 satisfies $|z(t)| \leq Z(t)$.

4. Since $L \in I_L$ and $|z(t)| \leq Z(t)$, we obtain the estimate for $t \in [-L_{min}, 0]$ below:

$$|z(t+L)| \le Z(t+L) \le \max_{L_{min} \le L \le L_{max}} Z(t+L) = Z_L(t).$$

Since by definition $z_L(t) = [z(t+L)]_0$, then for the definition of Z_L in Step 4, we have $|z_L(t)| \le Z_L(t)$ for $t \in [-1, 0]$.

5. Note that both y and x' satisfy (21), so by linearity z solves (21). Since $z_L(0) = 0$, we obtain the following estimate using variation of parameters:

$$z_L(-t) = -\int_{-t}^{0} -\alpha e^{x(s-1)} z_L(s-1) ds.$$

By taking the suprema over $\alpha \in I_{\alpha}$ and $\ell_K \leq x \leq u_K$ as in Step 5, we obtain a refinement for which $|z_L(t)| \leq Z_L(t)$. This refinement can be repeated any number of times.

6. If (λ, h) solves (21), then by (27) we obtain the following:

$$|\lambda| = \sup_{t \in [-1,0]} |z_L(t)| \le \sup_{t \in [-1,0]} Z_L(t) = \Lambda_{max}.$$

Hence $|\lambda| < \Lambda_{max}$ uniformly for $x \in \kappa^{-1}(K)$.

- 7. We have shown that $|\lambda| \leq \Lambda_{max}$ for any Floquet multiplier λ . If $\Lambda_{max} < 1$, then it follows that x is asymptotically stable.
- 8. If $\Lambda_{max} \geq 1$, then we make the assumption that x is not asymptotically stable for the sake of contradiction. Then the largest Floquet multiplier λ_{max} of x satisfies $|\lambda_{max}| \in [1, \Lambda_{max}]$. If h is an eigenfunction associated with λ_{max} , then $z_L(t) = \lambda_{max} \cdot h(t)$ for $t \in [-1, 0]$ and furthermore $|h(t)| = |\lambda_{max}|^{-1}|z_L(t)| \leq |z_L(t)|$. Hence for all $t \in [-1, 0]$ we may assume that the eigenfunction h(t) satisfies the inequality:

$$|h(t)| \le \min\{1, |z_L(t)|\}.$$

By definition y(t) = h(t) for $t \in [-1, 0]$. Hence for our refinement of Y in (25) it follows that $|y(t)| \le Y(t)$ for $t \in [-1, 0]$.

If the algorithm terminates having never passed through Step 8, then $|\lambda| \leq \Lambda_{max} < 1$ for all solutions (λ, h) to (22) uniformly for all SOPS $x \in \kappa^{-1}(K)$. If the program terminates having passed through Step 8 at least once, then it has shown that every solution (λ, h) to (22) satisfies $|\lambda| < 1$ under the assumption that there exists a solution for which $|\lambda| \geq 1$, a contradiction. In this case we have shown that x is asymptotically stable without calculating an explicit bound on its Floquet multipliers.

5 A Comprehensive Algorithm

We state our branch and prune algorithm in Algorithm 5.1, and describe how we use it to prove the uniqueness of SOPS to Wright's equation in Algorithm 5.3. Algorithm 5.1 takes as input an interval $I_{\alpha} \subseteq \mathbb{R}$ and constructs an I_{α} -exhaustive set. Furthermore, this algorithm uses several computational parameters: $\epsilon_1, \epsilon_2 \in \mathbb{R}$ which defines the algorithm's stopping criterion, $n_{Time} \in \mathbb{N}$ which defines the time resolution used in representing bounding functions on the computer, and $N_{prune} \in \mathbb{N}$ which defines the number of times the pruning algorithm is performed before branching. Additionally it requires the computational parameters $i_0, j_0, N_{Period} \in \mathbb{N}$ needed for running Algorithms 3.5 and 3.8. As we have stated before, this is a canonical algorithm which terminates in finite time (see [4, 15, 19]).

Algorithm 5.1. The input is an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and computational parameters $\epsilon_1, \epsilon_2 > 0$ and $i_0, j_0, n_{Time}, N_{Period}, N_{Prune} \in \mathbb{N}$. The output is a set $\mathcal{A} = \{K_i : K_i \subseteq \mathbb{R}^3\}$ and an associated collection of bounding functions $\{u_K, \ell_K\}_{K \in \mathcal{A}}$.

- 1. Construct regions K_1 and K_2 according to Algorithms 3.5 and 3.8 respectively. Define the sets $S = \{K_1, K_2\}$ and $A = \emptyset$.
- 2. If $S = \emptyset$ then return A and STOP.
- 3. Define K to be an element of S and remove K from S.
- 4. Define $\{K', u_{K'}, \ell_{K'}\}$ to be the output of Algorithm 2.2 using input K, u_K, ℓ_K and computational parameter n_{Time} . Then redefine $\{K, u_K, \ell_K\} \stackrel{\text{def}}{=} \{K', u_{K'}, \ell_{K'}\}$. Repeat this step N_{Prune} times.
- 5. If the diameter of K is less than ϵ_1 and $\bar{q} < 3$, or the diameter of K is less than ϵ_2 and $\bar{q} \geq 3$, then add K to A and GOTO Step 2.
- 6. Subdivide K along its fattest dimension into two regions K_A and K_B . That is, write $K = I_1 \times I_2 \times I_3$ where each I_i is given by the interval $I_i = [a_i, b_i]$ and fix some $j \in \{1, 2, 3\}$ which maximizes $|b_j a_j|$. The regions $K_A \stackrel{\text{def}}{=} I'_1 \times I'_2 \times I'_3$ and $K_B \stackrel{\text{def}}{=} I''_1 \times I''_2 \times I''_3$ are defined according to the following formulas

$$I_{i}^{\prime} \stackrel{\text{def}}{=} \begin{cases} [a_{i}, b_{i}] & \text{if } i \neq j \\ [a_{i}, (a_{i} + b_{i})/2] & \text{if } i = j \end{cases} \qquad I_{i}^{\prime\prime} \stackrel{\text{def}}{=} \begin{cases} [a_{i}, b_{i}] & \text{if } i \neq j \\ [(a_{i} + b_{i})/2, b_{i}] & \text{if } i = j. \end{cases}$$
 (28)

7. Add to S the regions K_A and K_B , each with associated bounding functions u_K and ℓ_K . Then GOTO Step 2.

As a notational convention for the next two theorems we define $\bigcup S \stackrel{\text{def}}{=} \bigcup_{K \in S} K$.

Theorem 5.2. Fix an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ such that $\alpha_{min} > \frac{\pi}{2}$, and fix any selection of computational parameters $\epsilon > 0$ and $i_0, j_0, n_{Time}, N_{Period}, N_{Prune} \in \mathbb{N}$. If \mathcal{A} is the output of Algorithm 5.1 with these inputs, then $\bigcup \mathcal{A}$ is I_{α} -exhaustive.

Proof. We prove by induction that every time the algorithm arrives at Step 2, then $\bigcup S \cup \bigcup A$ is I_{α} -exhaustive. This suffices to prove the theorem, as the only way for the algorithm to exit is on Line 2 when $S = \emptyset$.

For the initial case, the set $\bigcup S = K_1 \cup K_2$ produced in Step 1 is I_{α} -exhaustive by Proposition 3.12. The result of Step 3 simply rearranges the collection of regions, after which $\bigcup S \cup \bigcup A \cup K$ is I_{α} -exhaustive. In Step 4, this I_{α} -exhaustivity is maintained when replacing K with the output of Algorithm 2.2 as a direct result of Proposition 2.3. If Step 5 adds K to A, then when the algorithm arrives at Step 2 the set $\bigcup S \cup \bigcup A$ will be I_{α} -exhaustive. Otherwise Step 6 will divide K into two regions K_A and K_B for which $K = K_A \cup K_B$. Then in Step 7 both K_A and K_B are then added to S, after which $\bigcup S \cup \bigcup A$ is still I_{α} -exhaustive.

We are finally able to state our algorithm which can prove that Wright's equation has a unique SOPS over a given range of parameters.

Algorithm 5.3. The input is an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ and computational parameters $\epsilon_1, \epsilon_2 > 0$ and $i_0, j_0, n_{Time}, N_{Period}, N_{Prune}, N_{Floquet}, M_{Floquet} \in \mathbb{N}$. The output is a True or False statement.

- 1. Run Algorithm 5.1 with input I_{α} and computational parameters $\epsilon_1, \epsilon_2, i_0, j_0, n_{Time}, N_{Period}$ and N_{Prune} . Define A and $\{u_K, \ell_K\}_{K \in A}$ to be its output.
- 2. For each $K \in \mathcal{A}$ calculate $\Lambda_{max}(K)$ to be the output of Algorithm 4.2, run with input I_{α} , K, u_K , ℓ_K , and computational parameters n_{Time} , $N_{Floquet}$ and $M_{Floquet}$.
- 3. If $\Lambda_{max}(K) < 1$ for all $K \in \mathcal{A}$, then return TRUE. Otherwise return FALSE.

Theorem 5.4. Fix an interval $I_{\alpha} = [\alpha_{min}, \alpha_{max}]$ with $\alpha_{min} > \pi/2$. If Algorithm 5.3 returns the output TRUE for any selection of computational parameters $\epsilon > 0$, and $i_0, j_0, n_{Time}, N_{Prune}, N_{Floquet}, M_{Floquet} \in \mathbb{N}$, then there exists a unique SOPS to Wright's equation for all $\alpha \in I_{\alpha}$.

Proof. By Theorem 5.2 it follows that $\bigcup \mathcal{A} = \bigcup_{K \in \mathcal{A}} K$ is an I_{α} exhaustive set. That is, by Definition 2.1, up to a time translation any SOPS to Wright's equation for parameter $\alpha \in I_{\alpha}$ can be expressed as a function $x \in \mathcal{X}$ for which $\kappa(x) \in \bigcup \mathcal{A}$. If Algorithm 4.2 terminates with $\Lambda_{max}(K) < 1$ for all $K \in \mathcal{A}$, then by Theorem 4.3 it follows that any SOPS $x \in \mathcal{X}$ satisfying $\kappa(x) \in \bigcup \mathcal{A}$ must be asymptotically stable. Hence, by Theorem 1.5 it follows that there must be a unique SOPS to Wright's equation for each $\alpha \in I_{\alpha}$.

6 Discussion

In Algorithm 5.3 we defined an algorithm which, if successful, proves the uniqueness of SOPS to Wright's equation for a finite range of parameters I_{α} . Below we describe how we applied this algorithm to prove Theorem 1.3.

Proof of Theorem 1.3. To prove Theorem 1.3 we divide the interval [1.9, 6.0] into various subintervals I_{α} , and then divide each of these intervals into further subintervals of width $\Delta \alpha$. For example, the interval $I_{\alpha} = [2.1, 6.0]$ with $\Delta \alpha = 0.1$ was divided into subintervals [2.1, 2.2], [2.2, 2.3], ..., [5.9, 6.0]. The various computational parameters we used are given in the table below (see [28] for associated MATLAB code).

I_{lpha}	$\Delta \alpha$	n_{Time}	ϵ_1	ϵ_2	i_0	j_0	N_{Period}	N_{Prune}	$N_{Floquet}$	$M_{Floquet}$
[1.90, 1.96]	0.01	128	0.02	0.25	2	20	10	4	20	5
[1.96, 2.10]	0.01	64	0.05	0.25	2	20	10	4	20	5
[2.10, 6.00]	0.10	32	0.05	0.25	2	20	10	4	20	5

Table 1: For descriptions of how these parameters affect Algorithm 5.3, refer to Algorithms 3.5 and 3.8 for i_0 , j_0 and N_{Prune} ; refer to Algorithm 4.2 for $N_{Floquet}$ and $M_{Floquet}$; and refer to Algorithm 5.1 for ϵ_1 , ϵ_2 and N_{Prune} .

For each of these parameter values, we ran Algorithm 5.3 which returned TRUE as its output. By Theorem 5.4 it follows that there must be a unique SOPS to Wright's equation for each $\alpha \in [1.9, 6.0]$.

As described in Theorem 4.3, if Algorithm 4.2 terminates without having reached Step 8, then it produces explicit bounds on the Floquet multipliers of the SOPS to Wright's equation. These bounds are summarized in Figure 1. In the range [2.2, 6.0] Algorithm 4.2 exits on Step 7, so by Theorem 4.3 we obtain an upper bounds on the Floquet multipliers. In the regime $\alpha \in [1.90, 2.20]$ Algorithm 4.2 only terminated after reaching Step 8 at least once, so we are only able to deduce that any non-trivial Floquet multiplier has modulus strictly bounded above by 1. In total, the computation took 115 hours to run using a i7-5500U processor, and Algorithm 5.1 accounted for 94% of the computation time.

Running Algorithm 5.3 at high values of α is computationally expensive. This is because the period length of SOPS to Wright's equation grows exponentially [12], whereby our algorithm's run time and memory requirements also increases exponentially in α . Nevertheless, proving Theorem 1.3 with an upper limit of $\alpha = 6$ is sufficient for our purposes considering the results in [26] proved uniqueness for $\alpha > 5.67$.

A different challenge presents itself for decreasing the lower limit of $\alpha=1.9$ in Theorem 1.3. Namely, Xie's method for bounding the largest Floquet multiplier is not well suited to weakly attracting SOPS. Even when using precise numerical approximations (from [9]) of SOPS to Wright's equation at single values of α , Algorithm 4.2 was only able to show that the SOPS was asymptotically stable for values of α no lower than 1.85. By decreasing the parameters $\Delta \alpha$ and ϵ_1 , and increasing the other computational parameters, we could expect the uniqueness result for $\alpha \geq 1.9$ could be pushed closer to $\alpha = 1.85$. However we believe that new ideas are required in order to prove Conjecture 1.4.

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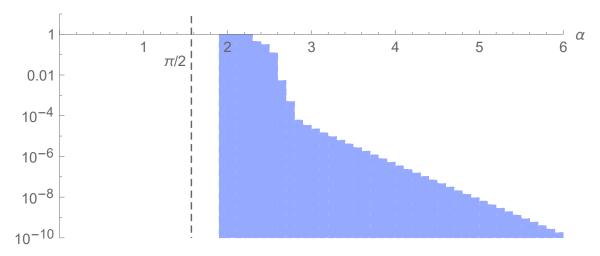


Figure 1: An upper bound on the modulus of the Floquet multipliers for SOPS to Wright's equation for $\alpha \in [1.9, 6.0]$.

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A Appendix: Computational Considerations

Interval arithmetic (e.g. see [11]) is an indispensable tool for producing computer-assisted proofs in nonlinear dynamics (e.g. see [8, 18, 22]). As the name suggests, interval arithmetic extends arithmetic operations to intervals of the real numbers, such as:

$$[a,b] + [c,d] \stackrel{\text{def}}{=} [a+c,b+d] \qquad \qquad [a,b] - [c,d] \stackrel{\text{def}}{=} [a-d,b-c].$$

In this manner, if $x_1 \in [a, b]$ and $x_2 \in [c, d]$ then $x_1 + x_2 \in [a, b] + [c, d]$. Moreover, for any function $f : \mathbb{R}^n \to \mathbb{R}$ and its interval extension F we have the relation:

$$f(x_1, x_2, \dots, x_n) \in F([a_1, b_1], [a_2, b_2], \dots, [a_n, b_n])$$

for all $x_i \in [a_i, b_i]$.

When interval arithmetic is performed on a computer the endpoints are necessarily binary numbers, and outward rounding is used to ensure proper enclosure. This allows us to compute a verified enclosure of the value of a function on numbers not representable on a binary computer, such as $\frac{1}{3}$. Furthermore this enables us to easily compute upper and lower bounds on a function over a rectangular domain of any size. While these bounds may not necessarily be sharp, they are guaranteed to be accurate.

To implement our algorithm we used Intlab: an interval arithmetic package for Matlab [17]. Some of the calculations we performed are a simple application of interval arithmetic, such as defining $I_q, I_{\bar{q}}, I_M$ in Algorithm 3.5. However there is a nontrivial degree of complexity in how we store and represent the functions used in the algorithms, such as u_K, ℓ_K in Algorithm 2.2 or Y, Z, Z_L in Algorithm 4.2. In a word, we defined these functions to be piecewise constant.

To explain our methodology, first fix a constant $n_{Time} \in \mathbb{N}$. To define an interval extension of a function $y : \mathbb{R} \to \mathbb{R}$, we define a collection of intervals $I_i^P, I_i^I \subseteq \mathbb{R}$ for $i \in \mathbb{Z}$ and define Y as follows:

$$Y(t) = \begin{cases} I_i^P & \text{if } t = \frac{i}{n_{Time}} \\ I_i^I & \text{if } t \in \left(\frac{i}{n_{Time}}, \frac{i+1}{n_{Time}}\right). \end{cases}$$

Of course any computer has finite memory, and so we would only store the function Y over a finite domain. Furthermore, as the bounding functions u, ℓ are intended to provide upper and lower bounds on a function x, we simply define an interval valued function $X(t) = [\ell(t), u(t)]$. In Figure 2 we present a graphical representation of how we store such a function, wherein we have defined the function X(t) for $t \in [-1, 0]$ as follows:

For such functions, it is a straightforward procedure to calculate its supremum. To calculate $\sup_{t \in [a,b]} x(t)$ one simply needs to compare the intervals I_i^P for which $a \leq \frac{i}{n_{Time}} \leq b$, and the intervals I_i^I for which $a - n_{Time}^{-1} < \frac{i}{n_{Time}} < b$. Both these collections of intervals are finite. For bounds which are defined to be the integrals of various functions, as in (15) and (16) of Lemma 3.7, we use a Riemann sum of step size $1/n_{Time}$.

Unfortunately there is a loss in fidelity when we numerically integrate these functions, as we do in Step 2 of Algorithm 2.2. Therein we refine the values of $u_{K'}(t_0+s)$, $\ell_{K'}(t_0+s)$, $u_{K'}(t_0-s)$ and $\ell_{K'}(t_0-s)$, where $t_0=\frac{i_0}{n_{Time}}$ and $s\in[0,\frac{1}{n_{Time}}]$. To just discuss the refinements of $u_{K'}(t_0+s)$ and $\ell_{K'}(t_0+s)$, if we choose $s=\frac{1}{n_{Time}}$, then this procedure refines the bound of $[\ell_{K'}(\frac{i_0+1}{n_{Time}}),u_{K'}(\frac{i_0+1}{n_{Time}})]$, a value which is stored in the interval $I_{i_0+1}^P$. However in order to refine $I_{i_0}^I$ this interval must include $[\ell_{K'}(t_0+s'),u_{K'}(t_0+s')]$ for all $s'\in(t_0,t_0+\frac{1}{n_{Time}})$. This is represented in Figure 2, where the darker red region represents the sharpest possible bounds able to be derived from in Step 2 of Algorithm 2.2 when integrating the initial data given above, and the pink region represents the values we store in the computer. When we define functions as integrals as in Steps 2 and 5 of Algorithm 4.2 we use the same procedure.

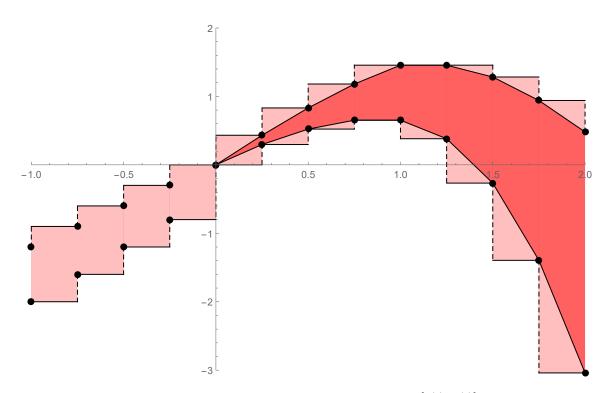


Figure 2: An example of how we store an interval valued function $[\ell(t), u(t)]$ in our algorithm.