

POINTWISE A POSTERIORI ERROR BOUNDS FOR BLOW-UP IN  
THE SEMILINEAR HEAT EQUATION\*IRENE KYZA<sup>†</sup> AND STEPHEN METCALFE<sup>‡</sup>

**Abstract.** This work is concerned with the development of an adaptive space-time numerical method, based on a rigorous a posteriori error bound, for the semilinear heat equation with a general local Lipschitz reaction term whose solution may blow up in finite time. More specifically, conditional a posteriori error bounds are derived in the  $L^\infty L^\infty$  norm for the first order (Euler) in time, implicit-explicit, conforming finite element method in space discretization of the problem. Numerical experiments applied to both blow-up and non-blow-up cases highlight the generality of our approach and complement the theoretical results.

**Key words.** semilinear heat equation, IMEX method, conditional a posteriori error estimates, blow-up singularities

**AMS subject classifications.** 65J08, 65L05, 65L60

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**1. Introduction.** Let  $\Omega \subset \mathbb{R}^d$  with  $d = 2$  or  $d = 3$  be a bounded polyhedral domain, and consider the problem

$$(1.1) \quad \begin{aligned} u_t - a\Delta u - f(u) &= 0 && \text{in } \Omega, \quad t > 0, \\ u &= 0 && \text{on } \partial\Omega, \quad t > 0, \\ u(\cdot, 0) &= u_0 && \text{in } \Omega, \end{aligned}$$

where  $a$  is a positive constant and the initial condition  $u_0 \in W^{2,\infty}(\Omega)$ . Note that here the reaction term  $f$  can be both space and time dependent, but as the nature of the dependence is usually clear, we omit writing it explicitly for brevity. It is well known that for certain data the solution to (1.1) exhibits *finite-time blow-up* [19], that is, there exists a maximal time of existence  $T_\infty < \infty$  referred to as the *blow-up time* such that (1.1) holds and

$$\|u(t)\|_{L^\infty(\Omega)} < \infty \text{ for } 0 < t < T_\infty, \quad \lim_{t \nearrow T_\infty} \|u(t)\|_{L^\infty(\Omega)} = \infty.$$

If the solution to (1.1) does not exhibit finite-time blow-up, then the solution is global and so  $T_\infty = \infty$ . Either way, we assume that (1.1) holds on some closed interval  $[0, T]$  and that  $T < T_\infty$ . We will see in what follows that we can show that (1.1) has a unique local solution  $u \in C(0, T; L^\infty(\Omega))$  provided that an implicit *local a posteriori criteron* is satisfied and that this local criteron is robust with respect to the distance from the blow-up time.

The numerical approximation of blow-up phenomena in partial differential equations (PDEs) is a challenging problem due to the high spatial and temporal resolution

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needed close to the blow-up time. In the literature, there exist several ad hoc numerical algorithms that produce good approximations to solutions of problems of the form (1.1) close to the blow-up time; see, for example, Acosta, Duran, and Rossi [1], Berger and Kohn [7], and Nguyen and Zaag [35]. Based on the moving mesh PDE method, Budd, Huang, and Russell also propose such algorithms (cf. [8, 20] and the references therein) while Groisman proposes an adaptive algorithm in [17] that is based on the asymptotic behavior of the exact solution. In all these works, information on the qualitative behavior of the exact solution close to the blow-up time is necessary; in fact, the aforementioned algorithms are driven by such behavior.

Recently, there has been a lot of interest [25, 9] in deriving rigorous a posteriori error bounds for problems of the form (1.1). Instead of relying on the qualitative behavior of the exact solution, the resulting estimators are used in order to design an efficient adaptive procedure that will allow us to get close to the blow-up time [9]. Indeed, it is easy to see why such an approach can confer significant advantages; for example, if it is known that the  $L^\infty L^\infty$  norm of the error is always bounded from above by a finite quantity, then it is impossible to surpass the blow-up time!

A posteriori error estimators for linear problems tend to be *unconditional*, that is, they always hold independent of the problem data and the size of the discretization parameters. For nonlinear problems, the situation is more complicated since the existence of a solution to an appropriate error equation (and, thus, of an error bound) usually requires that either the data or the discretization parameters are sufficiently small. As a result, a posteriori error estimators for nonlinear problems tend to be *conditional*, that is, they only hold provided that an a posteriori verifiable condition (which can be either explicit or implicit) is satisfied. For nonlinear time-dependent problems, there are two commonly used approaches for deriving conditional a posteriori error bounds: continuation arguments (cf. [6, 31, 26, 9, 21, 16, 11]) and fixed point arguments (cf. [25, 12, 24, 33, 34, 32]).

The derivation of such estimates for (1.1) and related problems in the context of blow-up was first explored in [24, 25] for polynomial nonlinearities, but these early pointwise bounds are not well suited for the practical computation of blow-up problems by virtue of being *global* rather than *local* in nature. The situation was improved in [31, 9] by the derivation of error bounds using energy techniques combined with the Gagliardo–Nirenberg inequality that are valid under local rather than global conditions. While the bounds of [31, 9] represent a significant improvement from a practical perspective, the derived error bounds still have significant drawbacks; for example, the range of nonlinearities that can be considered is smaller than in [24, 25] due to Sobolev embedding restrictions, and convergence towards the blow-up time is still *slow* when compared with results on the numerical approximation of blow-up in corresponding ODE cases [26, 31, 9]. It should be remarked, though, that the use of energy techniques in [31, 9] does confer advantages in other areas; specifically, it allows for the derivation of error bounds for problems with nonsymmetric spatial operators, e.g., for problems which include advective terms.

In this paper, we seek to derive conditional a posteriori error bounds in the  $L^\infty L^\infty$  norm for a fully discrete approximation of (1.1) where for the time discretization we use a first order (Euler) implicit-explicit (IMEX) method while for the spatial discretization we use the conforming finite element method. It is worth noting that the choice of an IMEX method not only confers advantages in terms of ease of solubility but also has been shown to have advantages in the context of estimation of the blow-up time; cf. [31, 9]. The results that we will present here are an improvement over existing results in the literature in two major ways. Firstly, we significantly broaden

the range of nonlinearities under consideration; specifically, the (possibly) nonlinear reaction term  $f : \bar{\Omega} \times [0, T_\infty) \times \mathbb{R} \rightarrow \mathbb{R}$  is assumed to be continuous and to satisfy the *local* Lipschitz estimate

$$(1.2) \quad |f(x, t, v) - f(x, t, w)| \leq \mathcal{L}(t, |v|, |w|)|v - w| \quad \forall x \in \bar{\Omega} \quad \forall t \in [0, T_\infty) \quad \forall v, w \in \mathbb{R}.$$

Here,  $\mathcal{L} : [0, T_\infty) \times \mathbb{R}_0^+ \times \mathbb{R}_0^+ \rightarrow \mathbb{R}_0^+$  is a *known* function that satisfies  $\mathcal{L}(\cdot, a, b) \in L^1(0, T_\infty)$  for any  $a, b \in \mathbb{R}_0^+$  and that is continuous and monotone increasing in the second and third arguments. This condition on  $f$  is quite general and includes many nonlinearities of interest; for example, it covers any polynomial nonlinearity with suitably regular coefficients as well as nonlinearities of exponential type [26]. We stress that this local Lipschitz assumption is in contrast to assumptions made for currently existing pointwise a posteriori bounds of (1.1) (cf. [22, 23]), wherein the focus is on the singularly perturbed case and, thus,  $f$  is assumed to be globally Lipschitz. Secondly, we follow the approach taken in [22, 24, 25, 23] of conducting the error analysis via semigroup techniques—this allows us to consider the error in an ODE setting. In combination with a *local-in-time* fixed point argument along the lines of [26], this restores the optimality that is otherwise lost in an energy setting [31, 9]. Additionally, we show numerically that our conditional a posteriori error bound is well behaved with respect to the distance from the blow-up time; specifically, we show that the rate of convergence to the blow-up time is comparable to the rate observed in [26, 9, 31]. Finally, we stress out that our error estimator is valid not only for problems with finite-time blow-up but for any problem of the form (1.1) satisfying (1.2)—we illustrate this in the last numerical experiment where we consider a problem without blow-up but with boundary layers instead.

**Outline.** We begin in section 2 by outlining the fully discrete IMEX finite element discretization of (1.1). Then in section 3, we introduce several auxiliary results which will be used in the error analysis, and we derive the conditional a posteriori error bound. In section 4, we propose a general adaptive algorithm, applicable to both blow-up and fixed-time problems (i.e., for problems without finite-time blow-up), that is based upon the derived a posteriori error bound. We apply this adaptive algorithm to several numerical examples in section 5 in order to illustrate that the proposed a posteriori bound is well behaved close to the blow-up time. Finally, we draw conclusions and outline our plans for future research in section 6.

**Notation.** As they will be used frequently in this work, we denote the  $L^2$  inner product on  $\Omega$  by  $(\cdot, \cdot)$  and the  $L^\infty$  norm on  $\Omega$  by  $\|\cdot\|$ .

**2. Discretization.** Consider a shape-regular mesh  $\mathcal{T} = \{K\}$  of  $\Omega$  with  $K$  denoting a generic element of diameter  $h_K$  that is constructed via affine mappings  $F_K : \hat{K} \rightarrow K$  with nonsingular Jacobian where  $\hat{K}$  is the  $d$ -dimensional reference simplex or the  $d$ -dimensional reference cube. The mesh is allowed to contain a uniformly fixed number of regular hanging nodes per face. With these definitions, the finite element space  $\mathcal{V}_h(\mathcal{T})$  over the mesh  $\mathcal{T}$  is given by

$$(2.1) \quad \mathcal{V}_h(\mathcal{T}) := \left\{ v \in H_0^1(\Omega) \cap C(\Omega) : v|_K \circ F_K \in \mathcal{P}^p(\hat{K}), K \in \mathcal{T} \right\},$$

where  $\mathcal{P}^p(\hat{K})$  denotes the space of polynomials of total degree  $p$  if  $\hat{K}$  is the  $d$ -dimensional reference simplex or of degree  $p$  in each variable if  $\hat{K}$  is the  $d$ -dimensional reference cube. Given two meshes  $\mathcal{T}_1$  and  $\mathcal{T}_2$ , we denote their *coarsest common refinement* by  $\mathcal{T}_1 \vee \mathcal{T}_2$  and their *finest common coarsening* by  $\mathcal{T}_1 \wedge \mathcal{T}_2$ . We also define the

jump residual  $\llbracket \nabla v_h \rrbracket$  of a function  $v_h \in \mathcal{V}_h(\mathcal{T})$  at a point  $x$  on the  $(d-1)$ -dimensional inter-element face  $E = \bar{K} \cap \bar{K}'$ ,  $K, K' \in \mathcal{T}$  by

$$\llbracket \nabla v_h \rrbracket(x) := \lim_{\delta \rightarrow 0} [\nabla v_h(x + \delta n) - \nabla v_h(x - \delta n)] \cdot n,$$

where  $n$  is an arbitrary unit normal vector on  $E$ .

We consider the first order in time, IMEX, space-time discretization of (1.1) consisting of implicit treatment for the diffusion term and explicit treatment for the nonlinear reaction term. For the spatial discretization, we use the standard conforming finite element method. To this end, we introduce a sequence of time nodes  $0 := t_0 < t_1 < \dots < t_{M-1} < t_M := T$  which induces a partition of  $[0, T]$  and let  $I_m := (t_{m-1}, t_m)$ ,  $m = 1, \dots, M$ . The length  $k_m := t_m - t_{m-1}$  (which may be variable) of the time interval  $I_m$  is called the time step length. If we let  $\mathcal{T}_1$  denote an initial spatial mesh of  $\Omega$  associated with the first time interval  $I_1$ , then to each successive time interval  $I_m$ ,  $m = 2, \dots, M$ , we associate a spatial mesh  $\mathcal{T}_m$  which is assumed to have been obtained from  $\mathcal{T}_{m-1}$  by local refinement and/or coarsening. We remark that this restriction upon mesh change is made in order to avoid degradation of the finite element solution; cf. [5, 15]. To each mesh  $\mathcal{T}_m$  we then assign the finite element space  $\mathcal{V}_h^m := \mathcal{V}_h(\mathcal{T}_m)$ ,  $m = 1, \dots, M$ , given by (2.1).

With this notation at hand, the IMEX fully discrete finite element method reads as follows: we seek  $U^m \in \mathcal{V}_h^m$ ,  $m = 1, \dots, M$ , such that

$$(2.2) \quad \left( \frac{U^m - U^{m-1}}{k_m}, V^m \right) + a(\nabla U^m, \nabla V^m) - (f^{m-1}, V^m) = 0 \quad \forall V^m \in \mathcal{V}_h^m$$

holds where we set  $f^{m-1} := f(\cdot, t_{m-1}, U^{m-1})$  for brevity and the initial value is given by  $U^0 := \mathcal{P}^1 u_0$  with  $\mathcal{P}^1$  denoting the  $L^2$ -projection operator onto  $\mathcal{V}_h^1$ . For  $t \in \bar{I}_m$ ,  $m = 1, \dots, M$ ,  $U(t)$  is then defined to be the linear interpolant with respect to  $t$  of the values  $U^{m-1}$  and  $U^m$ , viz.,

$$U(t) := \ell_{m-1}(t)U^{m-1} + \ell_m(t)U^m,$$

where  $\{\ell_{m-1}, \ell_m\}$  denotes the standard linear Lagrange interpolation basis on the interval  $\bar{I}_m$ , i.e.,  $\ell_{m-1}(t) := \frac{t_m - t}{k_m}$  and  $\ell_m(t) := \frac{t - t_{m-1}}{k_m}$ ,  $t \in \bar{I}_m$ .

**3. Error analysis.** Before we proceed with the error analysis, we require some auxiliary results and some additional notation. Our first result is a maximum principle for a related parabolic equation.

**THEOREM 3.1.** *Let  $e^{t\Delta}$  be the solution operator for the problem*

$$\begin{aligned} w_t - a\Delta w &= 0 && \text{in } \Omega, t > 0, \\ w &= 0 && \text{on } \partial\Omega, t > 0, \\ w(\cdot, 0) &= w_0 && \text{in } \Omega, \end{aligned}$$

*with  $w_0 \in L^\infty(\Omega)$ . In other words,  $w(t) = e^{t\Delta}w_0$ . Then for any  $t > 0$ , the following bound holds:*

$$\|e^{t\Delta}w_0\| \leq \|w_0\|.$$

*Proof.* See page 93 in [37]. □

We next introduce an error bound for a related elliptic problem which will be crucial in the error analysis of the parabolic problem.

**THEOREM 3.2.** *Let  $w \in H_0^1(\Omega) \cap C(\bar{\Omega})$  be the unique solution to the elliptic problem*

$$\begin{aligned} -a\Delta w &= g && \text{in } \Omega, \\ w &= 0 && \text{on } \partial\Omega, \end{aligned}$$

where  $g \in C(\Omega)$ , and let  $w_h \in \mathcal{V}_h$  be its conforming finite element approximation. Then the following pointwise *a posteriori* bound holds:

$$\|w - w_h\| \leq C_\infty \log(1/\underline{h}) \max_{K \in \mathcal{T}} [h_K^2 a^{-1} \|g + a\Delta w_h\|_{L^\infty(K)} + h_K \|\nabla w_h\|_{L^\infty(\partial K \setminus \partial\Omega)}],$$

where  $\underline{h} := \min_{K \in \mathcal{T}} h_K$  is the minimum mesh-size and  $C_\infty$  is a positive constant that is independent of the maximum mesh-size,  $a$ ,  $w$ , and  $w_h$  but may be dependent upon the size of the domain  $\Omega$ .

*Proof.* See [13]. □

**Remark 3.1.** The error bound of Theorem 3.2 and, thus, the spatial error estimators in the forthcoming parabolic error analysis are well behaved in the elliptic regime but badly behaved in the singularly perturbed regime ( $a \approx 0$ ); cf. [13]. As we are primarily interested in blow-up in the elliptic regime, we elect to use the simpler error bound of Theorem 3.2, but if one is interested in the singularly perturbed regime, the spatial estimators should be replaced with the *full* estimator from [13].

In the forthcoming error analysis, on each time step  $m$ , we will work with the Banach space  $C(\bar{I}_m; L^\infty(\Omega))$ —the space of all continuous functions  $v : \bar{I}_m \rightarrow L^\infty(\Omega)$  equipped with the norm

$$\|v\|_m := \sup_{t \in \bar{I}_m} \|v(t)\|.$$

In what follows, we will need to make use of the elliptic reconstruction technique originally introduced in [27, 28]. To that end, we define the *elliptic reconstruction*  $\omega^m \in H_0^1(\Omega)$ ,  $m = 1, \dots, M$ , to be the solution of

$$(3.1) \quad a(\nabla \omega^m, \nabla v) = (A^m, v) \quad \forall v \in H_0^1(\Omega),$$

where  $A^m := f^{m-1} - U_t|_{I_m}$ . Note that the  $L^2$ -projection of  $A^m$  onto  $\mathcal{V}_h^m$  is the discrete Laplacian of  $aU^m$ ; hence our definition of the elliptic reconstruction is not exactly the same as in [27, 28] but is instead in the spirit of [14]. Note that with this definition,  $A^m$  has nonzero boundary values. This modification is used to maintain consistency in the pointwise analysis [14] and leads to better numerical results in the implementation of the spatial *a posteriori* error estimator. For  $m = 0$ , we set  $\omega^0 := u_0$  and further define  $A^0 := -a\Delta u_0$ .

As with the numerical solution  $U(t)$ ,  $\omega(t)$  is defined to be the linear interpolant with respect to  $t$  of the values  $\omega^{m-1}$  and  $\omega^m$ , viz.,

$$(3.2) \quad \omega(t) := \ell_{m-1}(t)\omega^{m-1} + \ell_m(t)\omega^m,$$

for  $t \in \bar{I}_m$ ,  $m = 1, \dots, M$ . For the error analysis, the error  $e := u - U$  will be decomposed  $e = \rho + \epsilon$  where  $\rho := u - \omega$  is the *parabolic error* and  $\epsilon := \omega - U$  is the *elliptic error*. Note that for  $m = 0$  we have  $\rho(0) = 0$  and  $\epsilon(0) = e(0) = u_0 - \mathcal{P}^1 u_0$  (which is a computable quantity) whereas for  $m \geq 1$  we can estimate  $\|\epsilon(t_m)\|$  via elliptic error estimators currently available in the literature (based on the observation that (2.2) is the conforming finite element discretization of the elliptic problem (3.1)). To that end, we have the following lemma.

LEMMA 3.3. *For  $m \geq 1$ , the estimate*

$$\|\epsilon(t_m)\| \leq \mathcal{E}_m := C_\infty \log(1/h_m) \max_{K \in \mathcal{T}_m} \eta_S^m|_K$$

*holds with  $h_m := \min_{K \in \mathcal{T}_m} h_K$  denoting the minimum mesh-size,  $C_\infty$  the constant of Theorem 3.2, and where  $\eta_S^m$  is the primary space estimator given by*

$$\eta_S^m|_K := h_K^2 a^{-1} \|A^m + a\Delta U^m\|_{L^\infty(K)} + h_K \|\llbracket \nabla U^m \rrbracket\|_{L^\infty(\partial K \setminus \partial\Omega)}, \quad K \in \mathcal{T}_m.$$

*Proof.* Since  $A^m \in C(\Omega)$ , this follows directly from Theorem 3.2.  $\square$

To begin construction of the error equation, we first deduce from (3.1) and (3.2) that

$$(3.3) \quad -a\Delta\omega = \ell_{m-1} A^{m-1} + \ell_m A^m$$

for any  $t \in \bar{I}_m$ . We then subtract (3.3) from (1.1) to obtain

$$(3.4) \quad u_t - a\Delta\rho = f(u) - \ell_{m-1} A^{m-1} - \ell_m A^m.$$

Adding and subtracting  $f(\omega)$  and  $f(U)$  then yields

$$(3.5) \quad u_t - a\Delta\rho = f(\rho + \omega) - f(\omega) + f(\omega) - f(U) + f(U) - \ell_{m-1} A^{m-1} - \ell_m A^m.$$

Finally, adding and subtracting  $\omega_t$  and  $U_t$  gives the *error equation*

$$(3.6) \quad \rho_t - a\Delta\rho = f(\rho + \omega) - f(\omega) + f(\omega) - f(U) + R_T - \epsilon_t,$$

where  $R_T$  is the *temporal residual* given by  $R_T := f(U) - \ell_{m-1} A^{m-1} - \ell_m A^m - U_t$  in  $I_m$ . It can be seen that  $R_T$  is expected to be of optimal order in time by substituting  $A^{m-1}$  and  $A^m$ . Moreover, we note that  $R_T$  is a computable quantity and its accuracy can always be verified numerically; cf., e.g., Example 3 in section 6. Using the temporal residual, we then define the *time estimator*  $\eta_T^m$  on each time interval  $I_m$  by

$$\eta_T^m := \int_{I_m} \|R_T(s)\| \, ds.$$

**3.1. Fixed point argument.** We now seek to show that (3.6) has a unique solution  $\rho \in \mathcal{B}_m$  where  $\mathcal{B}_m$  is the closed ball of radius  $\delta_m \psi_m$  centered on zero in the  $\|\cdot\|_m$  norm and where  $\delta_m \in [1, \infty)$  is a parameter to be determined with  $\psi_m$  chosen such that

(3.7)

$$\|\rho(t_{m-1})\| + \int_{I_m} \|\epsilon(s)\| \mathcal{L}(s, \|U(s)\|, \|U(s)\| + \|\epsilon(s)\|) \, ds + \eta_T^m + \int_{I_m} \|\epsilon_t\| \, ds \leq \psi_m,$$

where  $\mathcal{L}$  is the local Lipschitz function from (1.2). The key idea here is that if we can show  $\rho$  exists in this ball, then  $\|\rho\|_m \leq \delta_m \psi_m$  were  $\delta_m \psi_m$  is our computable a posteriori error estimator. Here, the term  $\psi_m$  contains local-in-time residuals which are of optimal order in space and time as well as the error estimator from the previous time step while  $\delta_m$ , the solution of a nonlinear auxiliary equation, is a scaling parameter related to the nonlinearity which enforces the validity of the Banach fixed point theorem (cf. (3.10) and (3.11) below). To prove the existence of  $\rho \in \mathcal{B}_m$ , we will use the Banach fixed point theorem to show that  $\Phi_m$  given by

$$\begin{aligned}\Phi_m(v)(t) := & e^{(t-t_{m-1})\Delta} \rho(t_{m-1}) + \int_{t_{m-1}}^t e^{(t-s)\Delta} [f(v+\omega) - f(\omega) + f(\omega) - f(U)] ds \\ & + \int_{t_{m-1}}^t e^{(t-s)\Delta} [R_T - \epsilon_t] ds,\end{aligned}$$

$t \in \bar{I}_m$ , has a unique fixed point  $\rho \in \mathcal{B}_m$  which by Duhamel's principle must also solve (3.6). To satisfy the criteria of the Banach fixed point theorem, we must show two things:

- (I) That  $\Phi_m$  maps  $\mathcal{B}_m$  onto itself.
- (II) That  $\Phi_m$  is a contraction mapping.

We begin the verification of these criteria by applying Theorem 3.1 to  $\Phi_m$ . Thus for any  $t \in \bar{I}_m$  and  $v \in \mathcal{B}_m$  we have

$$\begin{aligned}\|\Phi_m(v)(t)\| \leq & \|\rho(t_{m-1})\| + \int_{t_{m-1}}^t \|f(v+\omega) - f(\omega)\| ds + \int_{I_m} \|f(\omega) - f(U)\| ds \\ & + \int_{I_m} \|R_T(s)\| ds + \int_{I_m} \|\epsilon_t\| ds.\end{aligned}$$

Using (1.2) with the monotonicity of  $\mathcal{L}$  and recalling (3.7) we obtain

$$(3.8) \quad \begin{aligned}\|\Phi_m(v)(t)\| \leq & \psi_m + \int_{t_{m-1}}^t \|v(s)\| \mathcal{L}(s, \|v(s)\| + \|U(s)\| + \|\epsilon(s)\|, \|U(s)\| + \|\epsilon(s)\|) ds.\end{aligned}$$

In order to bound this further, we require another lemma.

LEMMA 3.4. *For any  $t \in I_m$ , the estimate*

$$\|\epsilon(t)\| \leq \xi_m$$

holds where  $\xi_m := \max\{\mathcal{E}_{m-1}, \mathcal{E}_m\}$  with  $\mathcal{E}_0 := \|u_0 - \mathcal{P}^1 u_0\|$ .

*Proof.* Since  $\epsilon$  is linear in time, its maximal value occurs at either the left or right end point—the result then follows directly from Lemma 3.3 and upon recalling that  $\epsilon(0) = e(0)$ .  $\square$

Applying Lemma 3.4 to (3.8), using the monotonicity of  $\mathcal{L}$ , and noting that  $v \in \mathcal{B}_m$  yields

$$\|\Phi_m(v)\|_m \leq \psi_m \left[ 1 + \delta_m \int_{I_m} L(s, \delta_m) ds \right],$$

where

$$(3.9) \quad L(s, \delta) := \mathcal{L}(s, \delta \psi_m + \|U(s)\| + \xi_m, \delta \psi_m + \|U(s)\| + \xi_m), \quad s \in I_m, \delta \in [1, \infty).$$

Thus we obtain that property (I) is satisfied if

$$(3.10) \quad 1 + \delta_m \int_{I_m} L(s, \delta_m) ds \leq \delta_m.$$

To show that property (II) holds, let  $v_1, v_2 \in \mathcal{B}_m$ ; then the definition of  $\Phi_m$  implies that

$$(\Phi_m(v_1) - \Phi_m(v_2))(t) = \int_{t_{m-1}}^t e^{(t-s)\Delta} [f(v_1 + \omega) - f(v_2 + \omega)] ds, \quad t \in \bar{I}_m.$$

Applying Theorem 3.1 and Lemma 3.4 together with the monotonicity property of  $\mathcal{L}$  and noting that  $v_1, v_2 \in \mathcal{B}_m$  we obtain

$$\begin{aligned} & \|(\Phi_m(v_1) - \Phi_m(v_2))(t)\| \\ & \leq \int_{I_m} \|f(v_1 + \omega) - f(v_2 + \omega)\| ds \\ & \leq \int_{I_m} \|(v_1 - v_2)(s)\| \mathcal{L}(s, \|v_1(s)\| + \|U(s)\| + \|\epsilon(s)\|, \|v_2(s)\| + \|U(s)\| + \|\epsilon(s)\|) ds \\ & \leq \int_{I_m} L(s, \delta_m) \|(v_1 - v_2)(s)\| ds. \end{aligned}$$

Therefore,

$$\|\Phi_m(v_1) - \Phi_m(v_2)\|_m \leq \|v_1 - v_2\|_m \int_{I_m} L(s, \delta_m) ds,$$

which is a contraction if

$$(3.11) \quad \int_{I_m} L(s, \delta_m) ds < 1.$$

Note that (3.10) implies (3.11) since coupled with the fact that  $\delta_m \in [1, \infty)$ , (3.10) gives that

$$\int_{I_m} L(s, \delta_m) ds \leq 1 - \delta_m^{-1} < 1.$$

Thus if (3.10) is satisfied, then by the Banach fixed point theorem, (3.6) has a unique solution  $\rho \in \mathcal{B}_m$ . As we have *choice* over the value  $\delta_m$  can take in (3.10), then for practical reasons we choose  $\delta_m \in [1, \infty)$  to be, if it exists, the *smallest* root of the function  $\varphi_m : [1, \infty) \rightarrow \mathbb{R}$  defined by

$$(3.12) \quad \varphi_m(\delta) := 1 + \delta \left[ \int_{I_m} L(s, \delta) ds - 1 \right].$$

Furthermore, we can in fact obtain a tighter bound on  $\rho$  under no additional assumptions. Indeed, we now know that if  $\delta_m$  exists, then  $\Phi_m(\rho) = \rho \in \mathcal{B}_m$  satisfies (3.8); therefore, upon applying Gronwall's inequality and Lemma 3.4 to (3.8) we immediately deduce the bound

$$(3.13) \quad \|\rho\|_m \leq r_m \psi_m,$$

where  $r_m \geq 1$  is given by

$$(3.14) \quad r_m = \exp \left( \int_{I_m} \mathcal{L}(s, \delta_m \psi_m + \|U(s)\| + \xi_m, \|U(s)\| + \xi_m) ds \right).$$

Note that the upper bound in (3.13) is not yet computable as  $\psi_m$  is not yet an a posteriori quantity. We shall discuss this next.

*Remark 3.2.* With regards to blow-up problems, a natural question is how we expect the multiplicative factor  $r_m$  to scale. Firstly, we observe numerically that  $r_m \approx \delta_m$  for most time steps—this is not entirely unexpected since for  $\int_{I_m} L(s, \delta_m) ds$  small, (3.12) implies that

$$\delta_m = \frac{1}{1 - \int_{I_m} L(s, \delta_m) ds} \approx 1 + \int_{I_m} L(s, \delta_m) ds,$$

and so

$$r_m \approx 1 + \int_{I_m} \mathcal{L}(s, \delta_m \psi_m + \|U(s)\| + \xi_m, \|U(s)\| + \xi_m) ds \approx 1 + \int_{I_m} L(s, \delta_m) ds \approx \delta_m.$$

Then from numerical experiments for similar conditional ODE a posteriori error bounds [26] we would expect to observe that

$$\prod_{i=1}^m r_i \approx \prod_{i=1}^m \delta_i \propto \mathcal{F}(|T_\infty - t_m|^{-1})$$

for some function  $\mathcal{F} : \mathbb{R}_0^+ \rightarrow [1, \infty)$ ,  $\lim_{x \rightarrow \infty} \mathcal{F}(x) = \infty$  dependent upon the nonlinearity  $f$ . We will further investigate this numerically for quadratic nonlinearities in the Numerical Experiments section (cf., e.g., Figure 1 below).

**3.2. Computable error bound.** Now that  $\delta_m$  has been defined, we must characterize  $\psi_m$  from (3.7) in an a posteriori fashion in order to obtain a fully computable error bound. To do this, we must first estimate the term  $\|\rho(t_{m-1})\|$  along with the remaining terms containing  $\epsilon$  in (3.7).

In the previous subsection, we deduced that  $\rho \in \mathcal{B}_m$  if  $\delta_m$  exists; however, we assumed a priori knowledge of the existence of  $\rho(t_{m-1})$  in (3.7). To rectify this, we note that for  $m > 1$  we have

$$(3.15) \quad \|\rho(t_{m-1})\| \leq \|\rho\|_{m-1} \leq r_{m-1} \psi_{m-1}$$

if  $\rho \in \mathcal{B}_{m-1}$ . Similarly,  $\rho \in \mathcal{B}_{m-1}$  if  $\delta_{m-1}$  exists and if we can verify the existence of  $\rho(t_{m-2})$ . Continuing this argument inductively, we see that  $\rho \in \mathcal{B}_m$  provided that  $\delta_1, \dots, \delta_m$  exist.

With regards to the remaining terms in (3.7), the first term containing  $\epsilon$  can be estimated directly by using the monotonicity of  $\mathcal{L}$  combined with the bound of Lemma 3.4, viz.,

$$(3.16) \quad \int_{I_m} \|\epsilon(s)\| \mathcal{L}(s, \|U(s)\|, \|U(s)\| + \|\epsilon(s)\|) ds \leq \xi_m \int_{I_m} \mathcal{L}(s, \|U(s)\|, \|U(s)\| + \xi_m) ds.$$

To bound  $\epsilon_t$  term in (3.7), we require a lemma.

LEMMA 3.5. *The following bound holds:*

$$\int_{I_m} \|\epsilon_t\| ds \leq \xi'_m$$

with

$$\xi'_m := \begin{cases} \|u_0 - \mathcal{P}^1 u_0\| + C_\infty \log(1/h_m) \max_{K \in \mathcal{T}_m} \eta_S^m |_K & \text{if } m = 1, \\ C_\infty \log(1/\hat{h}_m) k_m \max_{\tilde{K} \in \mathcal{T}_{m-1} \wedge \mathcal{T}_m} \max_{\tilde{K} \subseteq \tilde{K}} \tilde{\eta}_S^m |_{\tilde{K}} & \text{if } m \geq 2, \end{cases}$$

where  $\widehat{h}_m := \min\{\underline{h}_{m-1}, h_m\}$  and where  $\dot{\eta}_S^m$  is the space derivative estimator given by

$$\begin{aligned}\dot{\eta}_S^m|_{\tilde{K}} &:= h_{\tilde{K}}^2 k_m^{-1} a^{-1} \|A^m - A^{m-1} + a\Delta(U^m - U^{m-1})\|_{L^\infty(\tilde{K})} \\ &\quad + h_{\tilde{K}} k_m^{-1} \|\nabla(U^m - U^{m-1})\|_{L^\infty(\partial\tilde{K} \setminus \partial\Omega)}\end{aligned}$$

for  $\mathcal{T}_{m-1} \vee \mathcal{T}_m \ni \tilde{K} \subseteq \widehat{K} \in \mathcal{T}_{m-1} \wedge \mathcal{T}_m$ . As before,  $C_\infty$  is a constant that is independent of the maximum mesh-size,  $a$ ,  $u$ , and  $U$  but may be dependent upon the size of the domain  $\Omega$  as well as the number of refinement levels between  $\mathcal{T}_{m-1}$  and  $\mathcal{T}_m$  for  $m \geq 2$ .

*Proof.* For  $m = 1$ , we have  $\epsilon_t = (\epsilon(t_1) - e(0))/k_1$ ; the stated result then follows from the triangle inequality and Lemma 3.3. For  $m \geq 2$ , we note the observation made in Corollary 2.9 of [14] that  $\epsilon_t$  is Galerkin orthogonal to the space  $\mathcal{V}_h^{m-1} \cap \mathcal{V}_h^m$ ; the stated bound then follows by conducting the error analysis as in [13] but with a quasi interpolant based on the mesh  $\mathcal{T}_{m-1} \wedge \mathcal{T}_m$ .  $\square$

Finally, applying the bounds of (3.15), (3.16) and Lemma 3.5 to the left-hand side of (3.7), we see that we can define  $\psi_m$  to be the computable a posteriori quantity given by

$$(3.17) \quad \psi_m := r_{m-1} \psi_{m-1} + \xi_m \int_{I_m} \mathcal{L}(s, \|U(s)\|, \|U(s)\| + \xi_m) \, ds + \eta_T^m + \xi'_m$$

with the initial values  $\psi_0 := 0$  and  $r_0 := 1$ . With  $\psi_m$  defined, all components of the error bound are now in place as well as fully computable, and we are ready to state the main result.

**THEOREM 3.6.** *Suppose that  $\delta_1, \dots, \delta_M$ , the smallest roots of (3.12) in  $[1, \infty)$ , exist; then the  $L^\infty L^\infty$  error of the IMEX method (2.2) satisfies the a posteriori error bound*

$$\max_{1 \leq m \leq M} \|e\|_m \leq r_M \psi_M + \max_{1 \leq m \leq M} \xi_m,$$

where  $r_M$  and  $\psi_M$  are defined (for  $m = M$ ) in (3.14) and (3.17), respectively, and  $\xi_m$  is given in Lemma 3.4.

*Proof.* Since  $\delta_1, \dots, \delta_M$  exist, by the exposition of the previous subsection we have

$$\|\rho\|_m \leq r_m \psi_m$$

for any  $1 \leq m \leq M$ . Noting the decomposition  $e = \rho + \epsilon$ , we obtain that

$$\max_{1 \leq m \leq M} \|e\|_m \leq \max_{1 \leq m \leq M} \|\rho\|_m + \max_{1 \leq m \leq M} \|\epsilon\|_m \leq r_M \psi_M + \max_{1 \leq m \leq M} \|\epsilon\|_m.$$

The stated result then follows from Lemma 3.4.  $\square$

Given these results, a natural question to ask is whether  $\delta_m \in [1, \infty)$ , the smallest root of  $\varphi_m$ , can actually exist at all. This is the focus of our next lemma.

**LEMMA 3.7.** *If  $\delta_0, \dots, \delta_{m-1}$  exist and the time step length  $k_m$  is chosen small enough, then  $\varphi_m$  of (3.12) has a root in  $[1, \infty)$ , and thus the a posteriori error bound of Theorem 3.6 holds on  $I_m$ .*

*Proof.* We proceed along the same lines as Lemma 3.1 in [26]. Since  $\delta_0, \dots, \delta_{m-1}$  exist, Theorem 3.6 implies that  $\psi_m$  exists and thus  $\varphi_m$  is well defined. For fixed

$\delta^* > 1$  and upon setting  $\epsilon = (\delta^* - 1)/2 > 0$  we can choose  $k_m > 0$  small enough so that

$$\delta^* \int_{I_m} L(s, \delta^*) ds < 1 + \epsilon.$$

A quick calculation reveals that  $\varphi_m(\delta^*) < 1 + \epsilon - \delta^* = (1 - \delta^*)/2 = -\epsilon < 0$ . Since  $\varphi_m$  is continuous and  $\varphi_m(1) > 0$ , it follows that for  $k_m$  sufficiently small  $\varphi_m$  has a root in  $[1, \infty)$ .  $\square$

We conclude this section by showing that the a posteriori error bound of Theorem 3.6 can be vastly simplified when  $f$  is independent of  $u$ .

**COROLLARY 3.8.** *Suppose that  $f$  is independent of  $u$ ; then the error of the IMEX method (2.2) unconditionally satisfies the a posteriori bound*

$$\max_{1 \leq m \leq M} \|e\|_m \leq \sum_{m=1}^M \eta_T^m + \sum_{m=1}^M \xi'_m + \max_{1 \leq m \leq M} \xi_m.$$

*Proof.* Since  $f$  is independent of  $u$ , it follows from (1.2) that  $\mathcal{L} = 0$ . We recall that  $\delta_m$  is the smallest root of the function  $\varphi_m : [1, \infty) \rightarrow \mathbb{R}$  given by

$$\varphi_m(\delta) = 1 + \delta \left[ \int_{I_m} L(s, \delta) ds - 1 \right] = 1 - \delta.$$

Therefore,  $\delta_m = 1$  regardless of the size of the time step length  $k_m$  and so the a posteriori error bound of Theorem 3.6 holds unconditionally. The stated result then follows from Theorem 3.6 by applying  $\psi_m$  recursively and upon noting that  $r_m = 1$ .  $\square$

**4. Adaptive algorithms.** In this section, we propose an adaptive algorithm that is based on the idea of minimization of the a posteriori error bound of Theorem 3.6. Ultimately, our goal is an adaptive algorithm that is applicable to both blow-up and fixed-time problems. Here, we consider an adaptive strategy that is based on using the residuals to control the time step lengths and mesh-sizes, but we emphasize that other adaptive strategies are possible for blow-up problems. For example, in [18], an existence analysis is carried out for implicit approximations to (1.1) via fixed point arguments, and the results are used to select the length of the time steps in the scheme. A similar adaptive strategy could be used here in the sense that we could continue to reduce the size of the time step on  $I_m$  until  $\delta_m$  exists and then fix this time step length before moving on to the next interval. Indeed, it was shown in [18] that choosing the size of the time steps in a way analogous to this can lead to superconvergence to the blow-up time; however, such a strategy does also come with certain disadvantages. Firstly, it is unclear how to generalize this approach to the case where  $f$  is independent of  $u$  since then  $\delta_m$  always exists. Moreover, it is unclear what a “natural” stop criterion for such an adaptive strategy should be whereas choosing the time step lengths according to the size of the residuals allows the nonexistence of  $\delta_m$  to be the stop criterion.

We contend that a general adaptive algorithm based on the a posteriori error bound of Theorem 3.6 should revert to a reasonable, well-known adaptive strategy when applied to a simple case such as that of Corollary 3.8. For this reason, we first outline an adaptive algorithm based on Corollary 3.8 and then extend this algorithm in a logical fashion to incorporate the full bound of Theorem 3.6. In what follows, we give a general outline of the idea behind the adaptive algorithms, but for brevity we also include the pseudocode of the full algorithm in Algorithm 1.

**Algorithm 1** Space-time adaptivity for the semilinear heat equation

---

```

1: Input:  $a, f, u_0, \Omega, T, \mathcal{T}_1, k_1, c_*, n, \text{stol}^+, \text{stol}^-, \text{ttol}^+, \text{ttol}^-$ .
2: Compute  $U^1$ .
3: while  $\text{ref}_T^1 > \text{ttol}^+$  or  $\max_{K \in \mathcal{T}_1} |\text{ref}_S^1|_K > c_* \text{stol}^+$  do
4:   Modify  $\mathcal{T}_1$  by refining all elements such that  $|\text{ref}_S^1|_K > c_* \text{stol}^+$  and coarsening
   all elements such that  $|\text{ref}_S^1|_K < c_* \text{stol}^-$ .
5:   if  $\text{ref}_T^1 > \text{ttol}^+$  then
6:      $k_1 \leftarrow k_1/2$ .
7:   end if
8:   Compute  $U^1$ .
9: end while
10: Attempt to compute  $\delta_1$ .
11: Set  $m = 0$ .
12: while  $\delta_{m+1}$  exists and  $t_{m+1} < T$  do
13:    $m \leftarrow m + 1$ .
14:   Set  $\mathcal{T}_{m+1} = \mathcal{T}_m$  and  $k_{m+1} = \min\{k_m, T - t_m\}$ .
15:   Compute  $U^{m+1}$  from  $U^m$ .
16:   if  $\text{ref}_T^{m+1} > \text{ttol}^+$  then
17:      $k_{m+1} \leftarrow k_{m+1}/2$ .
18:     Compute  $U^{m+1}$  from  $U^m$ .
19:   end if
20:   if  $\text{ref}_T^{m+1} < \text{ttol}^-$  then
21:      $k_{m+1} \leftarrow \min\{2k_{m+1}, T - t_m\}$ .
22:     Compute  $U^{m+1}$  from  $U^m$ .
23:   end if
24:   if  $m \bmod n = 0$  then
25:     Modify  $\mathcal{T}_{m+1}$  by refining all elements such that  $|\text{ref}_S^{m+1}|_K > \text{stol}^+$  and
        coarsening all elements such that  $|\text{ref}_S^{m+1}|_K < \text{stol}^-$ .
26:     Compute  $U^{m+1}$  from  $U^m$ .
27:   end if
28:   Attempt to compute  $\delta_{m+1}$ .
29: end while
30: if  $\delta_{m+1}$  exists and  $t_{m+1} = T$  then
31:    $m \leftarrow m + 1$ .
32: end if
33: Output:  $m, t_m, U, r_m \psi_m$ .

```

---

**4.1. Adaptive algorithm for Corollary 3.8.** The nature of the data in Corollary 3.8 means that the algorithm in this subsection will be for a *fixed-time problem*. As such, the inputs to the algorithm include the data, the domain  $\Omega$ , the final time  $T$ , a coarse initial mesh  $\mathcal{T}_1$  of  $\Omega$ , and an unrefined initial time step length  $k_1 \leq T$ .

Suppose that we are on the generic time step  $m > 1$ ; then the backward solution  $U^{m-1} \in \mathcal{V}_h^{m-1}$  and the mesh  $\mathcal{T}_{m-1}$  are already fixed. To proceed on the current interval, we first set the mesh  $\mathcal{T}_m = \mathcal{T}_{m-1}$  and then calculate the forward solution  $U^m \in \mathcal{V}_h^m$  given by (2.2). The simple structure of the error bound immediately suggests defining the refinement indicators

$$\text{ref}_T^m := \eta_T^m, \quad \text{ref}_S^m|_K := \max \left\{ \eta_S^m|_K, \dot{\eta}_S^m|_K \right\}, \quad K \in \mathcal{T}_m.$$

Note that  $\text{ref}_T^m$  is local to each time step while  $\text{ref}_S^m$  is local to each mesh element. As is standard for spatial adaptivity done via  $L^\infty$  norm error estimates, we ignore the global logarithmic terms [36]. To control the size of the time steps and the mesh elements, we introduce four tolerances: a *spatial refinement tolerance*  $\text{stol}^+$ , a *spatial coarsening tolerance*  $\text{stol}^-$ , a *temporal refinement tolerance*  $\text{ttol}^+$ , and a *temporal coarsening tolerance*  $\text{ttol}^-$ . If necessary, we begin by either refining or coarsening the time step length  $k_m$  and recalculating the forward solution  $U^m \in \mathcal{V}_h^m$  until

$$\text{ttol}^- \leq \text{ref}_T^m \leq \text{ttol}^+$$

is satisfied. We then fix this time step length. Next, we proceed spatially by, if permitted (see Remark 4.2 below), refining all elements  $K \in \mathcal{T}_m$  such that  $\text{ref}_S^m|_K > \text{stol}^+$  and coarsening all elements such that  $\text{ref}_S^m|_K < \text{stol}^-$ . We then recalculate (if necessary) and fix the forward solution  $U^m \in \mathcal{V}_h^m$ . After this is done, we set  $k_{m+1} = k_m$  and proceed to the next interval unless the total time  $t_{m+1} = t_m + k_{m+1}$  would surpass the final time  $T$  in which case we set  $k_{m+1} = T - t_m$ . When the final time is reached, we halt our computations.

All that remains is to deal with the coarse grid and time step length on the first interval. To do this, we must first modify the space refinement indicator in order to account for the unique form of  $\xi_m$  and  $\xi'_m$  on the first interval, viz.,

$$\text{ref}_S^1|_K := \max \left\{ \|u_0 - \mathcal{P}^1 u_0\|_{L^\infty(K)}, \eta_S^1|_K \right\}, \quad K \in \mathcal{T}_1.$$

Unfortunately,  $\text{ref}_S^1$  is of a different form to  $\text{ref}_S^m$  for  $m > 1$  which can cause problems with spatial adaptivity; specifically, the mismatch between the spatial indicators can lead to excessive overrefinement between the first and second time steps. Our proposed solution to this issue is to introduce a constant  $c_* \leq 1$  such that

$$c_* \approx \max_{K \in \mathcal{T}_1(\equiv \mathcal{T}_2)} \frac{\text{ref}_S^1|_K}{\text{ref}_S^2|_K}.$$

This constant is used to weight the spatial tolerances on the first interval only, i.e., we refine elements  $K \in \mathcal{T}_1$  such that  $\text{ref}_S^1|_K > c_* \text{stol}^+$  and coarsen elements such that  $\text{ref}_S^1|_K < c_* \text{stol}^-$ . This ensures roughly equal parity between the spatial refinement indicators on all time steps.

To begin refinement of the coarse initial mesh/time step length, we first calculate the numerical solution  $U^1 \in \mathcal{V}_h^1$  given by (2.2); we then proceed, via the tolerance strategy outlined above, by refining (or coarsening) the time step length  $k_1$  and the mesh  $\mathcal{T}_1$  concurrently, then recalculating the numerical solution  $U^1 \in \mathcal{V}_h^1$  until both  $\text{ref}_T^1 \leq \text{ttol}^+$  and  $\max_{K \in \mathcal{T}_1} \text{ref}_S^1|_K \leq c_* \text{stol}^+$  are satisfied. After this is done, we set  $k_2 = k_1$  and proceed to the next interval.

*Remark 4.1.* The refinement and coarsening tolerances need to be chosen sufficiently far apart so that the finite element solution does not get caught in an infinite refine and coarsen loop.

*Remark 4.2.* Numerical experiments show that the space derivative estimator  $\dot{\eta}_S^m$  becomes temporarily unstable after mesh change has occurred rendering its use in the spatial refinement indicator  $\text{ref}_S^m$  problematic. There are two practical ways to address this issue and prevent runaway degradation of the finite element method solution: we can remove  $\dot{\eta}_S^m$  from the spatial refinement indicator altogether, or we can

restrict mesh change to every few time steps in order to allow  $\dot{\eta}_S^m$  time to normalize. We opt for the second approach and only allow mesh change every  $n$  time steps as we consider this more desirable than the alternative which would involve excluding potential solution information from the spatial refinement indicator.

**4.2. Adaptive algorithm for Theorem 3.6.** In this subsection, we modify the adaptive algorithm of the previous subsection for adaptivity under the full estimator of Theorem 3.6; we do this in such a way that the algorithm of the previous subsection is recovered when  $f$  is independent of  $u$ . In fact, the only changes that need to be made to the algorithm of the previous subsection are the following:

(1) The stop criteron must be altered to account for the fact that the error bound may now not necessarily hold.

(2) The refinement indicators must be modified for adaptivity under the full error estimator.

To address point (1), we recall from Theorem 3.6 that the error bound holds on  $I_m$  provided that  $\delta_1, \dots, \delta_m$  exist; therefore, after all the adaptive procedures on the current interval are complete, we attempt to calculate  $\delta_m$  via a root-finding algorithm. If we find a root, then we continue to the next interval (unless the final time is reached); if not, we halt our computations.

A naïve approach towards addressing point (2) would be to simply use the refinement indicators of the previous subsection for adaptivity, however, doing so completely ignores not only the structure of the error estimator but also the length scales inherent to a blow-up problem. Consequently, such a refinement indicator causes excessive overrefinement close to the blow-up time which, in turn, results in suboptimal convergence [9, 31, 26]. In order to characterize the new refinement indicators, we define

$$\tilde{r}_m := \prod_{i=0}^m r_i$$

to be the accumulation of the values  $r_i$ ,  $0 \leq i \leq m$ . We remark that  $\tilde{r}_m$  is a value intimately connected with the rate of blow-up of the exact solution on the interval  $(0, t_m)$  as shown in the numerical experiments of [26] and as we shall show here in what follows; therefore, it is natural that this quantity should appear in our refinement indicators.

With this notation at hand, the temporal refinement indicator is given by

$$\text{ref}_T^m := \tilde{r}_{m-1}^{-1} \eta_T^m.$$

For a rigorous justification of this choice, we refer the reader to [26]. Defining the space refinement indicator is more involved as, reviewing the error bound of Theorem 3.6, it is clear that some terms are affected by  $r_m$  while others are not. The lone spatial term outside the recursive portion of the error bound is independent of  $r_m$  which suggests that we demand that

$$(4.1) \quad \eta_S^m|_K \leq \text{stol}^+, \quad K \in \mathcal{T}_m,$$

is satisfied. On the other hand, the space derivative estimator  $\dot{\eta}_S^m$  is part of the term  $\psi_m$  which is affected by  $r_m$ , and so we ask that

$$(4.2) \quad \tilde{r}_{m-1}^{-1} \dot{\eta}_S^m|_K \leq \text{stol}^+, \quad K \in \mathcal{T}_m,$$

is satisfied. The remaining spatial term in  $\psi_m$  is affected by  $r_m$  as well, but we must also divide it by the time step length  $k_m$  in order to incorporate it into the space refinement indicator as the space refinement indicator shouldn't be (strongly) dependent upon the time step length. Therefore, we require that

$$(4.3) \quad \left[ \tilde{r}_{m-1}^{-1} k_m^{-1} \int_{I_m} \mathcal{L}(s, \|U(s)\|, \|U(s)\| + \xi_m) ds \right] \eta_S^m |_K \leq \texttt{stol}^+, \quad K \in \mathcal{T}_m,$$

is satisfied. So upon defining

$$\alpha_m := \max \left\{ 1, \tilde{r}_{m-1}^{-1} k_m^{-1} \int_{I_m} \mathcal{L}(s, \|U(s)\|, \|U(s)\| + \xi_m) ds \right\},$$

we combine (4.1), (4.2), and (4.3) to define the space refinement indicator (for  $m > 1$ ), viz.,

$$\texttt{ref}_S^m |_K := \max \left\{ \alpha_m \eta_S^m |_K, \tilde{r}_{m-1}^{-1} \dot{\eta}_S^m |_K \right\}, \quad K \in \mathcal{T}_m.$$

As in the previous subsection, we must again modify the space refinement indicator on the first interval in order to account for the unique form of  $\xi_m$  and  $\xi'_m$  on that interval, viz.,

$$\texttt{ref}_S^1 |_K := \max \left\{ \alpha_1 \|u_0 - \mathcal{P}_1 u_0\|_{L^\infty(K)}, \alpha_1 \eta_S^1 |_K \right\}, \quad K \in \mathcal{T}_1.$$

Finally, we recall from Corollary 3.8 that for  $f$  independent of  $u$  we have  $\mathcal{L} = 0$  and  $\delta_m = 1$ . As  $\delta_m$  always exists for this type of data, the additional stop critereon introduced in this subsection never comes into play. Moreover,  $\mathcal{L} = 0$  implies  $\tilde{r}_m = 1$ , and so the reference indicators introduced here for the general case devolve into those of the previous subsection for  $f$  independent of  $u$ . Therefore, we conclude that the two adaptive algorithms are the same for this type of data.

**5. Numerical experiments.** We consider an implementation of the adaptive algorithm of the previous section through an application of the `deal.II` finite element library [4, 3]. For all of the numerical experiments, we only allow the mesh to change every three time steps. In order to facilitate a comparison between the  $L^\infty L^\infty$  estimator of Theorem 3.6 and the  $L^2 H^1$  estimator of [9], we consider Example 1 and Example 3 of [9] but under the adaptive algorithm of the previous section driven by the  $L^\infty L^\infty$  a posteriori error bound derived in this paper. If the a posteriori error bound of Theorem 3.6 is robust with respect to the distance from the blow-up time, then for sufficiently small  $\texttt{stol}^+$  and  $\texttt{stol}^-$ , we would expect to observe that

$$|T_\infty - T(\texttt{ttol}^+, N)| \propto N^{-1},$$

where  $T_\infty$  is the blow-up time of (1.1) and  $T$  is the final time produced by the adaptive algorithm in  $N$  total time steps under a given temporal refinement tolerance  $\texttt{ttol}^+$  as this is what was observed in the ODE experiments of [9, 26, 31]. Additionally, we also apply the adaptive algorithm to a nonlinear fixed-time problem in order to demonstrate its generality and to show that the estimator of Theorem 3.6 is of optimal order in space and time.

**5.1. Example 1.** Let  $\Omega = (-8, 8)^2$ ,  $a = 1$ ,  $f(u) = u^2$ , and choose the initial condition to be the Gaussian blob given by  $u_0(x, y) = 10 \exp(-2x^2 - 2y^2)$ . The blow-up set for this example consists of only a single point (the origin) making it spatially uncomplicated which allows us to focus solely on the temporal asymptotics. Now, since  $f(u) = u^2$ , for any  $v_1, v_2 \in \mathbb{R}$  we have

$$|f(v_1) - f(v_2)| = |v_1^2 - v_2^2| \leq |v_1 - v_2|(|v_1| + |v_2|).$$

Therefore, we have  $\mathcal{L}(|v_1|, |v_2|) = |v_1| + |v_2|$  in (1.2), and so  $\delta_m$  (if it exists) is the smallest root of the function  $\varphi_m : [1, \infty) \rightarrow \mathbb{R}$  given by

$$\begin{aligned} (5.1) \quad \varphi_m(\delta) &= 1 + \delta \left[ \int_{I_m} L(s, \delta) \, ds - 1 \right] \\ &= 1 + \delta \left[ 2k_m \xi_m + 2 \int_{I_m} \|U(s)\| \, ds - 1 \right] + 2k_m \psi_m \delta^2. \end{aligned}$$

In this case, we can calculate  $\delta_m$  explicitly via the quadratic formula, and so there is no need to use a root finding algorithm here.

Given that we are primarily interested in observing temporal behavior, we opt to use high degree polynomials (degree nine specifically) to minimize the spatial error. We begin by first setting a small spatial refinement tolerance  $\text{stol}^+$  so that the spatial error is negligible; we then gradually reduce the temporal refinement tolerance  $\text{ttol}^+$  in order to observe the rate of convergence to the blow-up time. We include the results in the left-hand side of Table 1 alongside the results that utilize the  $L^2 H^1$  estimator of [9] on the right-hand side.

The results show that for this example the  $L^\infty L^\infty$  estimator of Theorem 3.6 outperforms the  $L^2 H^1$  estimator of [9] in terms of rate of convergence to the blow-up time. We recall from [9] that given two consecutive data points we can approximate the exact blow-up time  $T_\infty$  as follows:

$$(5.2) \quad T_\infty \approx \frac{t_m \|U^m\| - t_{m-1} \|U^{m-1}\|}{\|U^m\| - \|U^{m-1}\|}.$$

Applying this here, we obtain the approximation  $T_\infty \approx 0.217015$ . Using this approximation to  $T_\infty$ , we take the data from Table 1 and plot the distance from the blow-up

TABLE 1  
Example 1:  $L^\infty L^\infty$  estimator of Theorem 3.6 (left) and  $L^2 H^1$  estimator of [9] (right).

$\text{ttol}^+$	Time steps	Final time	$\ U(T)\ $	Time steps	Final time	$\ U(T)\ $
				3	0.09375	12.244
0.25	2	0.05375	11.045	8	0.12500	14.742
0.25 <sup>2</sup>	9	0.12094	14.611	19	0.14844	18.556
0.25 <sup>3</sup>	23	0.15453	19.941	42	0.16406	23.468
0.25 <sup>4</sup>	52	0.17469	27.729	92	0.17969	32.108
0.25 <sup>5</sup>	114	0.19148	42.965	195	0.19043	44.217
0.25 <sup>6</sup>	397	0.20366	77.976	405	0.19775	60.493
0.25 <sup>7</sup>	816	0.20933	130.275	832	0.20313	83.315
0.25 <sup>8</sup>	1655	0.21227	208.52	1698	0.20728	117.780
0.25 <sup>9</sup>	3339	0.21421	347.669	3443	0.21014	165.833
0.25 <sup>10</sup>	6711	0.21537	584.525	6956	0.21228	238.705
0.25 <sup>11</sup>	13460	0.21606	998.835	14008	0.21375	343.078
				28151	0.21478	496.885
				56489	0.21549	722.884

time  $|T - T_\infty|$  versus the total number of time steps  $N$  in Figure 1. The plot shows that for this example we have

$$|T_\infty - T(\text{ttol}^+, N)| \propto N^{-3/4}.$$

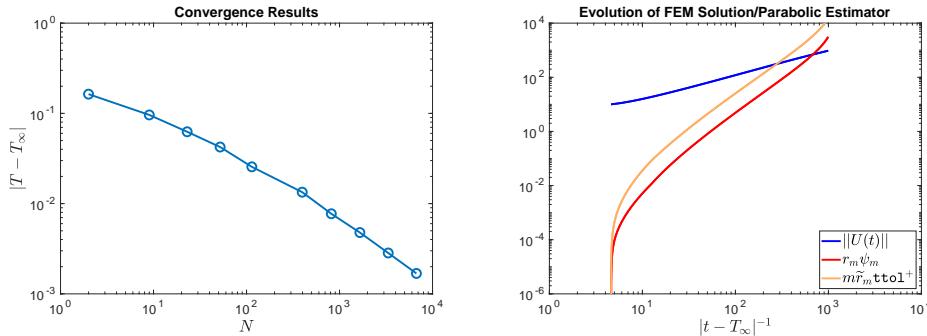


FIG. 1. Example 1: Convergence results (left) and evolution of the numerical solution (right).

This is slightly slower than expected given the ODE results of [9, 26, 31]. Finally, for the final computational run we plot the magnitude of the numerical solution  $\|U(t)\|$ , the parabolic estimator  $r_m \psi_m$ , and the value  $m \tilde{r}_m \text{ttol}^+$  versus the inverse of the distance to the blow-up time. From the results, given in Figure 1, we deduce the asymptotic estimate

$$\|U(t)\| \propto |t - T_\infty|^{-1},$$

which is consistent with the asymptotics of the exact solution [29, 30] suggesting that our numerical solution is reasonable. Moreover, we expect (cf. Corollary 4.3 of [26]) that, in the spatially asymptotic regime and under the adaptive strategy induced by the adaptive algorithm, the parabolic error  $\rho$  satisfies

$$\max_{1 \leq k \leq m} \|\rho\|_k \leq r_m \psi_m \leq m \tilde{r}_m \text{ttol}^+,$$

which we confirm in Figure 1. Therefore, upon observing that the gradient of the estimator curves in Figure 1 is two, we deduce that there exists a constant  $C > 0$  that is independent of the distance to the blow-up time such that

$$\max_{1 \leq k \leq m} \|\rho\|_k \leq C |t_m - T_\infty|^{-2}.$$

**5.2. Example 2.** Let  $\Omega = (-8, 8)^2$ ,  $a = 1$ ,  $f(u) = u^2$ , and the ‘‘volcano’’-type initial condition be given by  $u_0(x, y) = 10(x^2 + y^2) \exp(-0.5x^2 - 0.5y^2)$ . The blow-up set for this example is a circle centered on the origin making this example a good test of the spatial capabilities of the adaptive algorithm as many degrees of freedom are required in order to resolve the one-dimensional singularity close to the blow-up time. We remark that as the nonlinearity here is the same as in Example 1,  $\delta_m$  is again the smallest root of (5.1).

For this example, we again opt to use polynomials of degree nine in order to observe the temporal asymptotics. We proceed as in the previous example by choosing a small spatial refinement tolerance  $\text{stol}^+$  so that the spatial error is negligible; we then gradually reduce the temporal refinement tolerance  $\text{ttol}^+$  in order to observe

the rate of convergence to the blow-up time. The results, displayed on the left-hand side of Table 2 alongside the results that utilize the  $L^2H^1$  estimator of [9] on the right-hand side, show that the  $L^\infty L^\infty$  estimator of Theorem 3.6 again outperforms the  $L^2H^1$  estimator of [9] by an order of magnitude.

As the nonlinearity is the same here as in Example 1, (5.2) is still valid, and so we obtain the approximation  $T_\infty \approx 0.166460$ . Using this approximation to  $T_\infty$ , we take the data from Table 2 and plot the distance from the blow-up time  $|T - T_\infty|$  versus the total number of time steps  $N$  in Figure 2. The plot shows that for this example we have

$$|T_\infty - T(\text{ttol}^+, N)| \propto N^{-1},$$

which is what we expected to observe given the ODE results of [9, 26, 31]. Next, we investigate the spatial properties of the adaptive algorithm during the final computational run. We begin by plotting the number of degrees of freedom versus the inverse of the distance to the blow-up time in Figure 2. The plot shows a general (nonexcessive) increase in the number of degrees of freedom as we advance towards the blow-up time with some local decreases. To investigate this further, we display the meshes at times  $t = 0$  and  $t = T$  in Figure 3 which shows heavy refinement has occurred around the blow-up set with some derefinement in areas from which the finite element solution has retreated. For visualization purposes, we also display profile views of the finite element solution at times  $t = 0$  and  $t = T$  in Figure 3.

TABLE 2  
Example 2:  $L^\infty L^\infty$  estimator of Theorem 3.6 (left) and  $L^2H^1$  estimator of [9] (right).

<b>ttol<sup>+</sup></b>	Time steps	Final time	$\ U(T)\ $	Time steps	Final time	$\ U(T)\ $
0.25	4	0.08280	12.005	3	0.06250	10.371
0.25 <sup>2</sup>	11	0.11385	17.687	10	0.09375	14.194
0.25 <sup>3</sup>	26	0.13455	27.548	36	0.11979	21.842
0.25 <sup>4</sup>	71	0.15525	65.557	86	0.13412	31.446
0.25 <sup>5</sup>	159	0.16043	119.261	190	0.14388	45.122
0.25 <sup>6</sup>	333	0.16334	221.442	404	0.15072	64.907
0.25 <sup>7</sup>	714	0.16512	476.916	880	0.15601	98.048
0.25 <sup>8</sup>	1470	0.16572	881.210	1853	0.15942	146.162
0.25 <sup>9</sup>	2978	0.16607	1648.860	3831	0.16176	219.423
0.25 <sup>10</sup>	6120	0.16628	3508.360	7851	0.16336	332.849
0.25 <sup>11</sup>	12336	0.16637	6663.990	16137	0.16442	505.236
0.25 <sup>12</sup>	24938	0.16641	13241.900	32846	0.16512	769.652
				66442	0.16558	1175.210

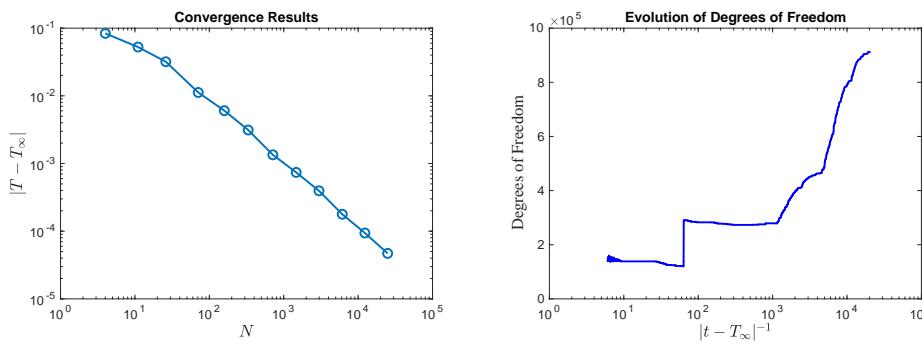


FIG. 2. Example 2: Convergence results (left) and evolution of degrees of freedom (right).

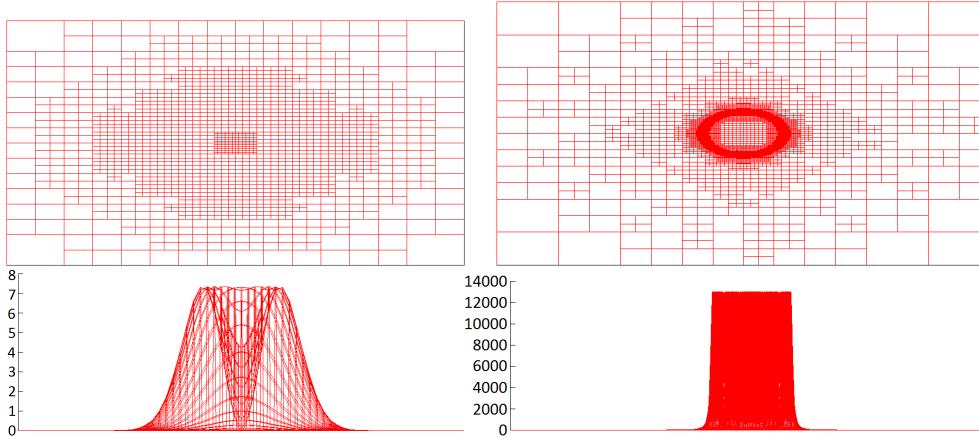


FIG. 3. Example 2: Initial mesh (top left), final mesh (top right), initial solution profile (bottom left), and final solution profile (bottom right).

**5.3. Example 3.** In this example, we consider a nonlinear parabolic problem from [2, 38]. We set  $\Omega = (0, 1)^2$ ,  $T = 0.5$ ,  $a = 0.001$ ,  $f(t, u) = \sin(t) - u^4$ , and  $u_0(x, y) = xy(x - 1)(y - 1)$ . The solution is initially unremarkable, but as time advances it begins to exhibit boundary layers through the influence of the diffusion and the forcing term. For this nonlinearity, given any  $t \in [0, T]$  and  $v_1, v_2 \in \mathbb{R}$  we have

$$|f(t, v_1) - f(t, v_2)| = |v_1^4 - v_2^4| \leq |v_1 - v_2|(|v_1|^3 + |v_1|^2|v_2| + |v_1||v_2|^2 + |v_2|^3).$$

Therefore, we have  $\mathcal{L}(|v_1|, |v_2|) = |v_1|^3 + |v_1|^2|v_2| + |v_1||v_2|^2 + |v_2|^3$  in (1.2), and so  $\delta_m$  (if it exists) is the smallest root of the function  $\varphi_m : [1, \infty) \rightarrow \mathbb{R}$  given by

$$\varphi_m(\delta) = 1 + \delta \left[ \int_{I_m} L(s, \delta) ds - 1 \right] = 1 - \delta + 4\delta \int_{I_m} (\delta \psi_m + \|U(s)\| + \xi_m)^3 ds,$$

which we approximate via a Newton method.

Our primary goal in this numerical example is to verify that the estimator of Theorem 3.6 is of optimal order in space and time when applied to a fixed-time problem. To that end, we begin by first checking the rate of convergence of the estimator in time. To do this, we choose polynomials of degree nine and a small spatial refinement tolerance `stol+` so that the spatial contribution to the estimator is negligible; we then gradually reduce the temporal refinement tolerance `ttol+` in order to observe the rate of convergence of the estimator in time. For each computational run, we then plot the value of the estimator at final time versus the total number of time steps. The results, displayed in Figure 4, show that the estimator is order one in time and, hence, optimal.

To analyze the rate of convergence of the estimator in space, we need to introduce a concept from [10] which is that of the *weighted average degrees of freedom* given by

$$\text{weighted average dofs} := \frac{1}{T} \sum_{m=1}^M k_m \times \# \text{dofs}(\mathcal{T}_m).$$

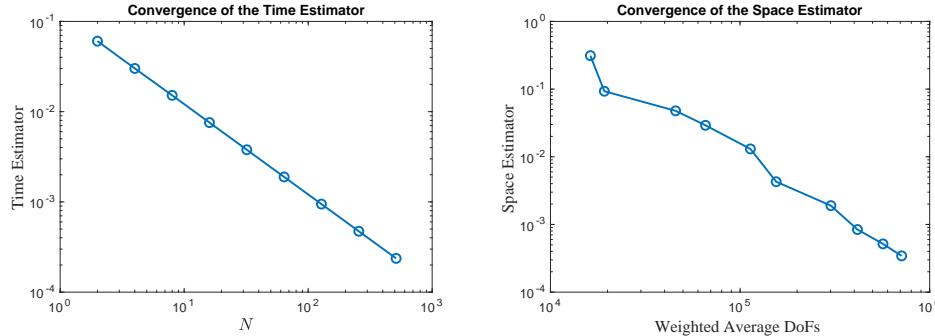


FIG. 4. Example 3: Temporal convergence (left) and spatial convergence for  $p = 3$  (right).

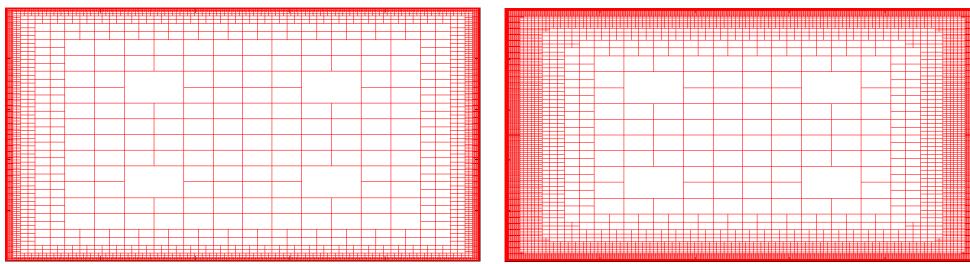


FIG. 5. Example 3: Initial mesh (left) and final mesh (right).

In order to observe the rate of convergence of the estimator in space, we first choose a small temporal refinement tolerance  $\text{ttol}^+$  so that the size of the temporal contribution to the estimator is negligible; we then gradually reduce the spatial refinement tolerance  $\text{stol}^+$  for polynomials of degree three and, for each computation, plot the value of the estimator at final time versus the weighted average degrees of freedom. The results, given in Figure 4, show that the spatial convergence is quite nonlinear but that we obtain the expected, optimal rate of convergence when the rate is calculated over all the computational runs. We also display the meshes from one of the computational runs at times  $t = 0$  and  $t = T$  in Figure 5. The initial mesh displays refinement around the boundary suggesting that the spatial estimator has correctly anticipated the formation of boundary layers. In the final mesh, we see that there is much heavier refinement around the boundary than in the initial mesh implying that the spatial estimator has accurately captured the evolution of the boundary layers as they formed.

**6. Conclusions.** We derived a conditional  $L^\infty L^\infty$  a posteriori error bound (Theorem 3.6) for the IMEX discretization (2.2) of the semilinear heat equation (1.1) with the general local Lipschitz nonlinearity (1.2). Our numerical experiments indicate that the proposed estimator outperforms the  $L^2 H^1$  estimator of [9] with respect to estimation of the blow-up time. Moreover, we were able to ascertain that the rate of convergence to the blow-up time is order one in the best-case scenario (Example 2), but we also determined that it can be slower (albeit still faster than in [9]) in certain situations (Example 1). The slow convergence in Example 1 can be explained by the initial condition not having a “compatible profile” with the blow-up, that is, this choice of initial condition causes the solution to be significantly influenced by the

Laplacian early on (this can be seen in Figure 1 by noting that the numerical solution only achieves the estimate  $\|U(t)\| \propto |t - T_\infty|^{-1}$  asymptotically and not at all stages of the computation) in a way that is not suitably accounted for by the proposed estimator. Attempts were made to modify the estimator of Theorem 3.6 to account for the influence of the Laplacian in the spirit of Proposition 4.5 of [14], but this did not appear to have a significant impact on the performance of the estimator. We note, however, that a requirement on the initial condition to have a “compatible profile” with the nonlinearity in order to achieve optimal convergence is not unreasonable and has been a requirement in the a posteriori error analysis of other nonlinear problems, for example, in [21]. Additionally, we verified in Example 3 that the estimator is of optimal order in space and time when applied to a fixed-time problem. Finally, we remark that condition (1.2) is very general and that the estimator of Theorem 3.6 can, in principle, be applied to *any* nonlinear problem which satisfies it. In practice, however, the exponential term  $\tilde{r}_m$  restricts application of the estimator to nonlinear problems for which either

- (1) the initial condition is small or
- (2) the nonlinearity is small or
- (3) the final time is small.

This also shows why the estimator works well for blow-up problems—because a large nonlinearity corresponds to a small blow-up time ensuring that  $\tilde{r}_m$  never grows out of control. In the future, we would like to robustly incorporate the influence of the Laplacian into the estimator of Theorem 3.6, prove convergence to the blow-up time under the proposed adaptive algorithm, and explore the possibility of exponential convergence to the blow-up time in the spirit of [26].

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#### REFERENCES

- [1] G. ACOSTA, R. G. DURÁN, AND J. D. ROSSI, *An adaptive time step procedure for a parabolic problem with blow-up*, Computing, 68 (2002), pp. 343–373.
- [2] M. AMREIN AND T. P. WIHLER, *An adaptive space-time Newton-Galerkin approach for semi-linear singularly perturbed parabolic evolution equations*, IMA J. Numer. Anal. 37 (2017), pp. 2004–2019.
- [3] W. BANGERTH, D. DAVYDOV, T. HEISTER, L. HELTAI, G. KANSCHAT, M. KRONBICHLER, M. MAIER, B. TURCKSIN, AND D. WELLS, *The deal.II library, version 8.4*, J. Numer. Math., 24 (2016), pp. 135–141.
- [4] W. BANGERTH, R. HARTMANN, AND G. KANSCHAT, *deal.II – a general-purpose object-oriented finite element library*, ACM Trans. Math. Software, 33 (2007), p. 24.
- [5] E. BÄNSCH, F. KARAKATSANI, AND C. MAKRIDAKIS, *The effect of mesh modification in time on the error control of fully discrete approximations for parabolic equations*, Appl. Numer. Math., 67 (2013), pp. 35–63.
- [6] S. BARTELS, *A posteriori error analysis for time-dependent Ginzburg-Landau type equations*, Numer. Math., 99 (2005), pp. 557–583.
- [7] M. BERGER AND R. V. KOHN, *A rescaling algorithm for the numerical calculation of blowing-up solutions*, Comm. Pure Appl. Math., 41 (1988), pp. 841–863.
- [8] C. J. BUDD, W. HUANG, AND R. D. RUSSELL, *Moving mesh methods for problems with blow-up*, SIAM J. Sci. Comput., 17 (1996), pp. 305–327.
- [9] A. CANGIANI, E. H. GEORGULIS, I. KYZA, AND S. METCALFE, *Adaptivity and blow-up detection for nonlinear evolution problems*, SIAM J. Sci. Comput., 38 (2016), pp. A3833–A3856, <https://doi.org/10.1137/16M106073X>.

- [10] A. CANGIANI, E. H. GEORGULIS, AND S. METCALFE, *Adaptive discontinuous Galerkin methods for nonstationary convection-diffusion problems*, IMA J. Numer. Anal. 34 (2014), pp. 1578–1597.
- [11] A. CANGIANI, E. H. GEORGULIS, A. Y. MOROZOV, AND O. J. SUTTON, *Revealing new dynamical patterns in a reaction-diffusion model with cyclic competition via a novel computational framework*, Proc. A, 474 (2018), 20170608.
- [12] E. CUESTA AND C. MAKRIDAKIS, *A posteriori error estimates and maximal regularity for approximations of fully nonlinear parabolic problems in Banach spaces*, Numer. Math., 110 (2008), pp. 257–275.
- [13] A. DEMLOW AND N. KOPTEVA, *Maximum-norm a posteriori error estimates for singularly perturbed elliptic reaction-diffusion problems*, Numer. Math., 133 (2016), pp. 707–742.
- [14] A. DEMLOW, O. LAKKIS, AND C. MAKRIDAKIS, *A posteriori error estimates in the maximum norm for parabolic problems*, SIAM J. Numer. Anal., 47 (2009), pp. 2157–2176.
- [15] T. DUPONT, *Mesh modification for evolution equations*, Math. Comp., 39 (1982), pp. 85–107, <https://doi.org/10.2307/2007621>.
- [16] E. H. GEORGULIS AND C. MAKRIDAKIS, *On a posteriori error control for the Allen-Cahn problem*, Math. Methods Appl. Sci., 37 (2014), pp. 173–179, <https://doi.org/10.1002/mma.2894>.
- [17] P. GROISMAN, *Totally discrete explicit and semi-implicit Euler methods for a blow-up problem in several space dimensions*, Computing, 76 (2006), pp. 325–352.
- [18] B. HOLM AND T. P. WIHLER, *Continuous and discontinuous Galerkin time stepping methods for nonlinear initial value problems with application to finite time blow-up*, Numer. Math., 138 (2018), pp. 767–799, <https://doi.org/10.1007/s00211-017-0918-2>.
- [19] B. HU, *Blow-Up Theories for Semilinear Parabolic Equations*, Lecture Notes in Math. 2018, Springer, Heidelberg, 2011.
- [20] W. HUANG, J. MA, AND R. D. RUSSELL, *A study of moving mesh PDE methods for numerical simulation of blowup in reaction diffusion equations*, J. Comput. Phys., 227 (2008), pp. 6532–6552.
- [21] D. KESSLER, R. H. NOCHETTO, AND A. SCHMIDT, *A posteriori error control for the Allen-Cahn problem: Circumventing Gronwall's inequality*, ESAIM Math. Model. Numer. Anal., 38 (2004), pp. 129–142, <http://eudml.org/doc/245771>.
- [22] N. KOPTEVA AND T. LINSS, *Maximum norm a posteriori error estimation for parabolic problems using elliptic reconstructions*, SIAM J. Numer. Anal., 51 (2016), pp. 1494–1524.
- [23] N. KOPTEVA AND T. LINSS, *Improved maximum-norm a posteriori error estimates for linear and semilinear parabolic equations*, Adv. Comput. Math., 43 (2017), pp. 999–1022, <https://doi.org/10.1007/s10444-017-9514-3>.
- [24] I. KYZA, *A Posteriori Error Estimates for Approximations of Semilinear Parabolic and Schrödinger-Type Equations*, PhD thesis, University of Crete, 2009.
- [25] I. KYZA AND C. MAKRIDAKIS, *Analysis for time discrete approximations of blow-up solutions of semilinear parabolic equations*, SIAM J. Numer. Anal., 49 (2011), pp. 405–426, <https://doi.org/10.1137/110846397>.
- [26] I. KYZA, S. METCALFE, AND T. P. WIHLER, *hp-adaptive Galerkin time stepping methods for nonlinear initial value problems*, J. Sci. Comput., 75 (2018), pp. 111–127, <https://doi.org/10.1007/s10915-017-0565-x>.
- [27] O. LAKKIS AND C. MAKRIDAKIS, *Elliptic reconstruction and a posteriori error estimates for fully discrete linear parabolic problems*, Math. Comp., 75 (2006), pp. 1627–1658, <https://doi.org/10.1090/S0025-5718-06-01858-8>.
- [28] C. MAKRIDAKIS AND R. H. NOCHETTO, *Elliptic reconstruction and a posteriori error estimates for parabolic problems*, SIAM J. Numer. Anal., 41 (2003), pp. 1585–1594, <https://doi.org/10.1137/S0036142902406314>.
- [29] F. MERLE AND H. ZAAG, *Optimal estimates for blowup rate and behavior for nonlinear heat equations*, Comm. Pure Appl. Math., 51 (1998), pp. 139–196.
- [30] F. MERLE AND H. ZAAG, *A Liouville theorem for vector-valued nonlinear heat equations and applications*, Math. Ann., 316 (2000), pp. 103–137.
- [31] S. METCALFE, *Adaptive Discontinuous Galerkin Methods for Nonlinear Parabolic Problems*, PhD thesis, University of Leicester, 2015.
- [32] M. MIZUGUCHI, A. TAKAYASU, T. KUBO, AND S. OISHI, *Accurate method of verified computing for solutions of semilinear heat equations*, Reliab. Comput., 25 (2017), pp. 74–99.
- [33] M. MIZUGUCHI, A. TAKAYASU, T. KUBO, AND S. OISHI, *A method of verified computations for solutions to semilinear parabolic equations using semigroup theory*, SIAM J. Numer. Anal., 55 (2017), pp. 980–1001.

- [34] M. MIZUGUCHI, A. TAKAYASU, T. KUBO, AND S. OISHI, *Numerical verification for existence of a global-in-time solution to semilinear parabolic equations*, J. Comput. Appl. Math., 315 (2017), pp. 1–16.
- [35] V. T. NGUYEN AND H. ZAAG, *Blow-up results for a strongly perturbed semilinear heat equation: Theoretical analysis and numerical method*, Anal. PDE, 9 (2016), pp. 229–257.
- [36] R. H. NOCHETTO, A. SCHMIDT, K. G. SIEBERT, AND A. VEESER, *Pointwise a posteriori error estimates for monotone semi-linear equations*, Numer. Math., 104 (2006), pp. 515–538.
- [37] V. THOMÉE, *Galerkin Finite Element Methods for Parabolic Problems*, Springer, Cham, 1984.
- [38] F. VERHULST, *Methods and Applications of Singular Perturbations: Boundary Layers and Multiple Timescale Dynamics*, Springer Science & Business Media, New York, 2005.