Abstract. We propose and analyze an algorithm for the solution of the $L^2$-subgradient flow of the total variation functional. The algorithm involves no regularization, thus the numerical solution preserves the main features that motivate practitioners to consider this type of energy. We propose an iterative scheme for the solution of the arising problems, show that the iterations converge, and develop a stopping criterion for them. We present numerical experiments which illustrate the power of the method, explore the solution behavior, and compare with regularized flows.

Key words. Total Variation; Singular Diffusion; Maximal Monotone Operators; Subgradient Flows; Variational Inequalities.

AMS subject classifications. 65N12; 65M60; 65N30; 35K86; 35R35; 35R37; 65K15.

1. Introduction. This work is concerned with the approximation of solutions to the following (formal) initial boundary value problem:

$$
\begin{aligned}
\frac{\partial u}{\partial t} &= \nabla \cdot \left( \frac{\nabla u}{|\nabla u|} \right), & \text{in } \Omega \times (0, T), \\
\frac{\nabla u}{|\nabla u|} \cdot n &= 0, & \text{on } \partial \Omega, \\
u|_{t=0} &= u_0, & \text{in } \Omega.
\end{aligned}
$$

Here $\Omega$ is an open, bounded and connected subset of $\mathbb{R}^d$ with $d \geq 1$; $\partial \Omega$ denotes the boundary of $\Omega$ and $n$ its exterior unit normal; and $T > 0$ is a positive and finite time. This equation and related ones are commonly known as very singular diffusion equations (see [26, 27]) since in flat regions, i.e., $|\nabla u| = 0$, the diffusion is so strong that it is not a local effect anymore.

Beginning with the seminal paper [34] equations of this class have received considerable attention from the image processing community, since such models preserve discontinuities while removing noise and other artifacts (see also [11, 13, 16, 17, 31]). In addition, such equations appear in the modeling of grain boundary motion [29]; facet formation and evolution [23]; electromigration [25] and various other problems stemming from materials science. This clearly shows that the development of efficient and accurate numerical schemes for the solution of this class of problems is of extreme importance. To the best of our knowledge however, the techniques advocated in the literature for the solution of these equations as a rule involve a regularization somewhat related with replacing the singular term by

$$
|\nabla u|_\epsilon = \sqrt{\epsilon^2 + |\nabla u|^2}, \quad \epsilon > 0;
$$

see, for instance, [9, 19, 21, 22, 30]. The disadvantages of this approach are twofold: first, although these methods have been shown to converge, there is no clear understanding of the relation between the regularization parameter $\epsilon$ and the discretization...
parameters; second, the regularization destroys certain fundamental features of the
solutions which motivate the introduction of such models in the first place.

A notable exception to the trend mentioned above is the work [7] which, inspired
by the ideas advanced in [12], develops a finite element scheme for total variation
minimization which involves no regularization. In this work we adapt and extend the
ideas presented in [7] to the study of total variation flows. We propose and analyze
an unconditionally stable and convergent discretization scheme for the approximation
of (1.1). In addition, we study an iterative scheme for the solution of the discrete
problems and develop an a posteriori error estimator that provides a robust stop-
ping criterion for the iterative scheme that guarantees that, although we only have
approximate solutions, the convergence properties of the method are not affected.

This work is organized as follows. The notation and conventions are set in Sec-
tion 2. In §2.1 we recall the definition and main properties of functions of bounded
variation and, in addition, we present an approximation result for these functions.
This will be our workhorse during the derivation of error estimates. The proper frame-
work to understand (1.1), i.e., total variation flow and its properties are described in
Section 3. Time and space discretization are discussed in Section 4, where we begin
by reviewing results on implicit semidiscretizations of gradient flows in Hilbert spaces.
Then we provide a general theory for fully discrete subgradient flows in Hilbert spaces,
which we later apply to our problem. One of the salient novelties of this general the-
ory is that we allow for modifications in the energy which can be used, for instance,
to take into account the effects of quadrature. In Section 5 we propose an iterative
scheme for the solution of the fully discrete problems, and devise a stopping criterion
for the iterations, which guarantees that the convergence orders are not damaged.
Section 6 contains a series of numerical experiments, which illustrate and extend the
properties and theory for the developed scheme. Finally, in Section 7, we apply the
approximation result of §2.1 to the problem of total variation minimization and prove
convergence estimates similar to those available in the literature, but with reduced
regularity assumptions.

2. Notation and Preliminaries. We denote by $\Omega$ a bounded domain in $\mathbb{R}^d$
with $d \geq 1$. The boundary of $\Omega$ is denoted by $\partial \Omega$ and we assume that $\partial \Omega \in C^{0,1}$. We
set $T > 0$ to be a finite final time. As usual, we denote by $L^p(\Omega)$ the space of Lebesgue
integrable functions with exponent $p \in [1, \infty]$ and by $W^s_p(\Omega), s \in \mathbb{R}$, the usual Sobolev
spaces. Spaces of vector valued functions and their elements will be represented with
boldface characters. Recall that the following interpolation inequality holds (cf. [24])
\[ \|w\|_{L^r} \leq \|w\|_{L^p}^{1-s} \|w\|_{L^q}^s, \quad p, q \in [1, \infty], \quad \frac{1}{r} = \frac{s}{p} + \frac{1-s}{q}, \quad s \in [0,1]. \] (2.1)

Whenever $E$ is a normed space, we denote its norm by $\| \cdot \|_E$ and its dual by
$E'$. The $L^2$-inner product will be denoted by $\langle \cdot, \cdot \rangle$. Function spaces of vector-valued
functions will be denoted by boldface characters. For function spaces, if it is clear
from the context, we will omit the domain of definition. We denote by $\mathcal{B}_1(E)$ the
unit ball in $E$, i.e., the set $\mathcal{B}_1(E) = \{ x \in E : \|x\|_E \leq 1 \}$. Let $\phi : [0, T] \to E$ be a
measurable function in the Bochner sense, then, for $p \in [1, \infty]$, we define
\[ \|\phi\|_{L^p(E)}^p = \int_0^T \|\phi(t)\|_E^p \, dt, \quad p < \infty, \quad \|\phi\|_{L^\infty(E)} = \text{ess sup}_{t \in [0, T]} \|\phi(t)\|_E. \]

We will denote by $\frac{d\phi}{dt}$ the time derivative of $\phi$. 

To deal with time discretization, we introduce a time-step $\Delta t > 0$, for simplicity assumed constant. Then we partition the time interval via $t^k = k\Delta t$ with $k = 0, K$ and $K = \lceil T/\Delta t \rceil$. We use the notation $\phi^{\Delta t} = \{ \phi^k \}_{k=0}^K$, introduce the time increment operator

$$\delta \phi^k = \phi^k - \phi^{k-1},$$  \hfill (2.2)

and the extrapolation operator

$$\phi^{*,k+1} = \phi^0, \quad k = 0, \quad \phi^{*,k+1} = \phi^k + \delta \phi^k, \quad k > 0.$$  \hfill (2.3)

For $\phi^{\Delta t} \subset E$ and $p \in [1, \infty]$, we introduce the (semi)norms

$$\| \phi^{\Delta t} \|_{L^p(E)}^p = \Delta t \sum_{k=0}^K \| \phi^k \|_{E}^p, \quad \| \phi^{\Delta t} \|_{L^\infty(E)}^2 = \sum_{k=1}^K \| \delta \phi^k \|_{E}^2,$$

with the usual modification for $p = \infty$. Given a sequence $\phi^{\Delta t} \subset E$ we will want to be able to compare it to functions defined on $[0,T]$. To this end, we define the Rothe interpolant $\hat{\phi}$, that is, the piecewise linear function

$$\hat{\phi}(t) = \frac{t - t^k}{\Delta t} \phi^k + \frac{t^{k+1} - t}{\Delta t} \phi^k, \quad t \in [t^k, t^{k+1}].$$  \hfill (2.4)

Notice that, by construction, $\hat{\phi} \in C^{0,1}(E)$, and $\| \hat{\phi} \|_{L^\infty(E)} = \| \phi^{\Delta t} \|_{L^\infty(E)}$. Recall also the well-known summation by parts formula: for every $\phi^{\Delta t}, \psi^{\Delta t} \subset L^2(\Omega)$,

$$\sum_{k=0}^{K-1} \left( \langle \phi^k, \delta \psi^{k+1} \rangle + \langle \delta \phi^{k+1}, \psi^k \rangle \right) = \langle \phi^K, \psi^K \rangle - \langle \phi^0, \psi^0 \rangle.$$  \hfill (2.5)

We will carry out the space discretization with finite element techniques. In other words, given $\Omega$ we introduce a so-called triangulation $T_h = \{ T \}$ as a collection of cells that satisfy the usual conformity and shape regularity assumptions [14, 20], and are such that $\Omega = \bigcup_{T \in T_h} \bar{T}$. We parametrize our collection of triangulations via $h = \max \{ \mathrm{diam}(T) : T \in T_h \}$. For simplicity, we assume that each $T$ is the isoparametric image of a so-called reference cell, which can be either $\bar{T} = [-1,1]^d$, in which case we call the cells cubic; or $\bar{T} = \left\{ (x_1, \ldots, x_d) \in \mathbb{R}^d : x_i \geq 0, \sum_{i=1}^d x_i \leq 1 \right\}$, which are called simplices. We denote by $\{ z_j, T \}$ the vertices of the cell $T$ and set $N_T = \# \{ z_j, T \}$. Clearly, $N_T = 2^d$ for cubes and $N_T = d + 1$ for simplices.

We define the finite element space

$$V_h = \{ w_h \in C^0(\Omega) : v_h|_T \in \mathcal{P} \} \subset W^1_{\infty}(\Omega),$$  \hfill (2.6)

where $\mathcal{P} = \mathbb{Q}_1$ for cubes and $\mathcal{P} = \mathbb{P}_1$ for simplices. Here $\mathbb{Q}_1$ denotes the space of polynomials of degree at most one in each variable and $\mathbb{P}_1$ the space of polynomials of total degree not greater than one.

For a function $w$ such that $w|_T \in C^0(\bar{T})$, we define its local Lagrange interpolant $I_h w$ by

$$I_h w|_T \in \mathcal{P} : I_h w|_T(z_j, T) = w|_T(z_j, T), \quad j = 1, N_T, \quad \forall T \in T_h.$$  \hfill (2.7)
This operator satisfies
\[ \| w - I_h w \|_{L^p(T)} + h \| \nabla (w - I_h w) \|_{L^p(T)} \leq c h^2 \| D^2 w \|_{L^p(T)}, \]  
(2.8)
see, e.g. [1], for a proof. We remark also that if \( w \in C^0(\Omega) \), then \( I_h w \in \mathcal{V}_h \). It will be also necessary to introduce the Clément interpolant (cf. [15]) \( \Pi_h : L^1(\Omega) \to \mathcal{V}_h \). This operator enjoys approximation properties similar to (2.8), the only difference being that the domain on the right hand side is a neighborhood of \( T \). More importantly, the operator is stable under any Sobolev norm, i.e.,
\[ \| \Pi_h w \|_{W^s_p} \leq c \| w \|_{W^s_p}, \quad s \geq 0, \quad p \in [1, \infty]. \]  
(2.9)

We will denote by \( c \) a constant whose value might change at each occurrence.

2.1. Functions of Bounded Variation and their Approximation. We say that a function \( w \in L^1(\Omega) \) belongs to the space \( BV(\Omega) \) (is of bounded variation) if its derivative, \( Dw \), in the sense of distributions is a Radon measure. In other words, \( |Dw|(\Omega) < \infty \), where for any Borel set \( A \subset \Omega \),
\[ |Dw|(A) = \sup \left\{ \int_A w \nabla q : q \in C^\infty_0(A), \| q \|_{L^\infty(A)} \leq 1 \right\}. \]
The space \( BV(\Omega) \) endowed with the norm \( \| w \|_{BV} = \| w \|_{L^1} + |Dw|(\Omega) \) is a Banach space. For more details on this space, we refer to [1] [38]. Let us present a result on approximation by smooth functions, in the case of a star-shaped domain.

**Proposition 2.1 (Approximation of BV functions).** Assume that \( \Omega \) is bounded, star-shaped with respect to a point and \( \partial \Omega \in C^{0,1} \). If \( w \in BV(\Omega) \), then for every \( \epsilon > 0 \) there exists a \( w_\epsilon \in C^\infty(\Omega) \) such that
\[ \| w - w_\epsilon \|_{L^1} \leq \epsilon |Dw|(\Omega), \quad \| \nabla w_\epsilon \|_{L^1} \leq (1 + \epsilon) |Dw|(\Omega), \quad \| D^2 w_\epsilon \|_{L^1} \leq c \epsilon^{-1} |Dw|(\Omega). \]

**Proof.** Without loss of generality, we can assume that \( 0 \in \Omega \) and that \( \Omega \) is star-shaped with respect to \( 0 \). For \(\epsilon > 0\) we define
\[ \Omega_\epsilon = \{ y \in \mathbb{R}^d : y = (1+\epsilon) x, \ x \in \Omega \}, \]
and notice that \( \Omega \) and \( \Omega_\epsilon \) are related via a bijective and Lipschitz, in fact linear, transformation with Lipschitz constant \( 1 + \epsilon \), and Jacobian \( 1 + \epsilon^d \leq 1 + c \epsilon \). For \( w \in BV(\Omega) \) we define \( v_\epsilon \in BV(\Omega_\epsilon) \) via \( v_\epsilon(y) = w(\frac{y}{1+\epsilon}) \) and, for \( x \in \Omega, \ w_\epsilon(x) = w_\epsilon \ast \rho_\epsilon(x) \), where \( \rho_\epsilon \) is a smooth convolution kernel such that, for ever \( 1 \leq p \leq \infty, \|
abla \rho_\epsilon \|_{L^p} \leq c \epsilon^{-(1+d/p')} \) with \( p' = p/(p-1) \).

If \( w \in C^1(\Omega) \) then, clearly, \( \| w - w_\epsilon \|_{L^1} \leq c \| \nabla w_\epsilon \|_{L^1} = c |Dw|(\Omega) \) and
\[ \| \nabla w_\epsilon \|_{L^1} \leq |Dv_\epsilon|(\Omega_\epsilon) \leq (1 + c \epsilon) \| \nabla w \|_{L^1} = (1 + c \epsilon) |Dw|(\Omega). \]

We now recall that smooth functions are dense in \( BV(\Omega) \) under strict convergence, (cf. [1] [38]). In other words, given \( w \in BV(\Omega) \) there is a sequence \( \{ w_n \}_{n \in \mathbb{N}} \subset C^\infty(\Omega) \) such that
\[ \lim_{n \to \infty} \| w - w_n \|_{L^1} = 0, \quad \limsup_{n \to \infty} |Dw_n|(\Omega) \leq |Dw|(\Omega). \]
Applying the argument given above to elements of this sequence and then passing to the limit we obtain the first two inequalities.
We use Young’s inequality for convolutions [24, Proposition 8.7] to obtain
\[ \|D^2 w_\epsilon\|_{L^1} \leq \|\nabla w_\epsilon\|_{L^1} \|\nabla \rho_\epsilon\|_{L^1} \leq c \frac{1 + c\epsilon}{\epsilon} |Dw|(|\Omega|) \leq \frac{c}{\epsilon} |Dw|(|\Omega|), \]
which concludes the proof.

**Remark 2.2 (Approximation in $L^p$-spaces).** Observe that, in the setting of Proposition 2.1, Young’s inequality also implies
\[ \|D^2 w_\epsilon\|_{L^p} \leq \|\nabla w_\epsilon\|_{L^1} \|\nabla \rho_\epsilon\|_{L^p} \leq c e^{-(1+d/p')}|Dw|(|\Omega|), \]
for any $p \in [1, \infty]$ with $p' = p/(p - 1)$. In the sequel, however, we shall avoid using this bound since it would lead to $d$-dependent error estimates.

### 3. The Total Variation Flow

Let us define the functional $\Psi : L^2(\Omega) \to \mathbb{R}$ by
\[ \Psi(w) = \begin{cases} |Dw|(\Omega), & w \in L^2(\Omega) \cap BV(\Omega), \\ +\infty, & w \in L^2(\Omega) \setminus BV(\Omega), \end{cases} \quad (3.1) \]
It is not difficult to show that $\Psi$ is convex and lower semicontinuous. Then, one can define the subdifferential of $\Psi$ (see [2, 6, 8, 10, 18, 26, 27]) and study its subgradient flow, i.e., we seek for a function $u : [0, T] \to L^2(\Omega)$ such that
\[ u_t \in -\partial \Psi(u), \quad (3.2) \]
or, equivalently,
\[ \langle u_t, u - w \rangle + \Psi(u) - \Psi(w) \leq 0, \quad \forall w \in L^2(\Omega). \quad (3.3) \]
It is in this sense that (1.1) is going to be understood and analyzed. Existence of solutions to (3.2) can be obtained with the help of the theory of maximal monotone operators, see [6, 10].

**Remark 3.1 (Dirichlet boundary conditions).** The definition that we have provided corresponds to imposing Neumann boundary conditions as in (1.1). The issue of how to impose Dirichlet boundary conditions is a delicate one since the trace of a $BV(\Omega)$ function is in $L^1(\partial \Omega)$ (cf. [4, 28]). In addition, the functional (3.1) has linear growth, so that the only possible way to impose Dirichlet boundary conditions is with the relaxed energy $\Psi(w) + \int_{\partial \Omega} |w - g|$: see [2, 28] for details. The introduction of this additional non-differentiable term greatly complicates the analysis. However, in the particular situation when $\Omega$ is convex, the boundary data is time independent and continuous, i.e., $g(x,t) = g(x) \in C^0(\partial \Omega)$, and the initial data is compatible with the boundary data in the sense that $u_0|_{\partial \Omega} = g$, then the solution $u$ to the subgradient flow with the relaxed energy satisfies $u(\cdot, t)|_{\partial \Omega} = g(\cdot)$, for all $t \in (0, T]$; see [28, Lemma 4.1]. This can be realized, for instance, in the case of homogeneous Dirichlet boundary conditions ($g \equiv 0$) and compactly supported initial data. Under this particular setting all the results we present will also hold.

Reference [4, Theorem 2.16] shows that problem (3.2) possesses a $L^d$-$L^\infty$ regularizing effect, that is if $u_0 \in L^d(\Omega)$, then $u(t) \in L^\infty(\Omega)$ for all $t > 0$. Let us show that solutions to this problem also satisfy a maximum principle.

**Theorem 3.2 (Maximum principle for TV flow).** Assume that $u_0 \in L^\infty(\Omega)$, then $u$, solution of (3.3), is such that $u \in L^\infty([0, T], L^\infty(\Omega))$ and
\[ \|u\|_{L^\infty(L^\infty)} \leq \|u_0\|_{L^\infty}. \]
Proof. Since $u_0 \in L^\infty(\Omega)$ we can define $U_0 = \text{ess sup}_{x \in \Omega} u_0(x)$, and $\tilde{w}(t) = u(t) \vee U_0$. Since $\text{supp}(D\tilde{w}) = \{x : u \geq U_0\}$ and, on this set, $D\tilde{w} = Du$, we can conclude that $\tilde{w} \in L^2(\Omega) \cap BV(\Omega)$ and $\Psi(\tilde{w}) \leq \Psi(u)$. Setting $w = \tilde{w}$ in (3.3) we obtain $\langle u_0, 0 \vee (u - U_0) \rangle \leq 0$, which, since $U_0$ is constant, implies $\frac{d}{dt}\|u - U_0\|_2^2 \leq 0$. Given that $(u_0 - U_0) \vee 0 = 0$ this implies the result. \(\square\)

4. Discretization. In this section we introduce and analyze a fully discrete scheme for the approximation of solutions to (3.3). We begin with a semidiscrete (continuous in space and discrete in time) scheme for (3.3) which can then be analyzed using standard results from the literature (cf. [33, 35]). Then we develop a theory for fully discrete subgradient flows in Hilbert spaces and discuss the effect of introducing a discrete energy and a perturbation on the right hand side. We will provide sufficient compatibility conditions between the space discretization and the discrete energy to guarantee convergence. These results constitute a general and, as far as we know, novel approach to the study of fully discrete schemes for subgradient flows and evolution variational inequalities. The main application of these results will be, of course, a fully discrete scheme for (3.3).

4.1. A Semidiscrete Scheme for TV Flows. We introduce a sequence $\{u^{\Delta t}\}$ contained in $L^2(\Omega) \cap BV(\Omega)$ with $u^0 = u_0$ that solves:

$$
\left\langle \frac{\delta u^{k+1}}{\Delta t}, u^{k+1} - w \right\rangle + \Psi(u^{k+1}) - \Psi(w) \leq 0 \quad \forall w \in L^2(\Omega). \tag{4.1}
$$

Existence and uniqueness is guaranteed by the convexity and lower semicontinuity of $\Psi$; see [2, 6, 10, 18]. A priori estimates, as well as a maximum principle are established in the next result.

**Proposition 4.1 (Semidiscrete stability).** Let $\{u^{\Delta t}\}$ solve (4.1). If $u^0 \in L^2(\Omega)$, then

$$
\|u^{\Delta t}\|_{L^2(\Omega)}^2 + \|u^{\Delta t}\|_{b^{1/2}(L^2)}^2 + \Delta t \sum_{k=1}^{K} |Du^k|(\Omega) \leq c\|u^0\|_{L^2}^2. \tag{4.2}
$$

If $u^0 \in L^2(\Omega) \cap BV(\Omega)$, then the flow is monotone, i.e., $\Psi(u^{k+1}) \leq \Psi(u^k) \leq \Psi(u^0)$ for all $k \geq 0$; and, moreover,

$$
\|u^{\Delta t}\|_{b^{1/2}(L^2)}^2 + \Delta t \Psi(u^K) \leq \Delta t \Psi(u^0). \tag{4.3}
$$

If $u^0 \in L^\infty(\Omega) \cap BV(\Omega)$, then $u^{\Delta t} \subset L^\infty(\Omega)$ and

$$
\|u^{\Delta t}\|_{L^\infty(\Omega)} \leq \|u^0\|_{L^\infty}. \tag{4.4}
$$

**Proof.** To obtain (4.2) it suffices to set $w = 0$ on (4.1) and add over $k$. To obtain (4.3) we set $w = u^k$ and add over $k$. The maximum principle is obtained mutatis mutandis the proof of Theorem 3.2 \(\square\)

The convergence properties of (4.1) are a consequence of standard results [33, 35]. For convenience we summarize them below.

**Corollary 4.2 (Convergence of semidiscrete TV flow).** Assume that $u^0 \in L^2(\Omega) \cap BV(\Omega)$, then

$$
\|u - \hat{u}\|_{L^\infty(\Omega)}^2 + \Delta t \Psi(u^K) \leq c \left(\|u_0 - u_0\|_{L^2}^2 + \Delta t^2\right),
$$

where $\hat{u}$ is the solution of (4.1).
where \( \hat{u} \) is the piecewise linear function defined in [2.4] and \( \alpha = 1 \). If, in addition, \( \partial \Psi(u_0) \cap L^2(\Omega) \neq \emptyset \), then \( \alpha = 2 \).

**Remark 4.3 (Error estimates for semidiscrete flows).** Notice that the conclusion of Corollary 4.2 gives an error estimate of order \( \mathcal{O}(\Delta t^{1/2}) \) under the sole assumption that the initial condition has finite energy. Under this regularity the result is optimal, see [33, Theorem 5]. Observe also that the assumption \( \partial \Psi(u_0) \cap L^2(\Omega) \neq \emptyset \), necessary to obtain a \( \mathcal{O}(\Delta t) \) estimate, is not a regularity but rather a compatibility assumption. We refer the reader to [32] for a partial characterization of the subdifferential of the total variation.

### 4.2. Fully Discrete Schemes for Subgradient Flows in Hilbert Spaces.

Here we present and analyze of a fully discrete implicit Euler method for subgradient flows in Hilbert spaces. The main novelty of our approach is the treatment of the space discretization and that we allow for perturbations of the energy, as well as of the right hand side.

Let \( H \) be a Hilbert space with inner product \( \langle \cdot, \cdot \rangle \), which induces the norm \( \| \cdot \|_H \). Assume that the functional \( \mathfrak{F} : D(\mathfrak{F}) \subset H \to \mathbb{R} \) is convex, lower semicontinuous and \( D(\mathfrak{F}) = H \). Then, for every \( w \in D(\mathfrak{F}) \) the subdifferential \( \partial \mathfrak{F}(w) \subset H \) is not empty. For details we refer to [6, 10, 18]. We want to study its subgradient flow: Find \( v : [0, T] \to H \) with \( v|_{t=0} = v_0 \), such that

\[
\dot{v}_t + \partial \mathfrak{F}(v) \ni 0, \quad (4.4)
\]
or, equivalently,

\[
\langle \dot{v}_t, v - w \rangle + \mathfrak{F}(v) - \mathfrak{F}(w) \leq 0, \quad \forall w \in H. \quad (4.5)
\]

The existence and uniqueness of a solution to (4.4) or (4.5) follows the theory of maximal monotone operators, [6, 10].

To discretize in time we consider the Euler method. In other words, we search for \( v^{\Delta t} \subset H \), \( v^0 = v_0 \), such that

\[
\left\langle \frac{\delta v^{k+1}}{\Delta t}, v^{k+1} - w \right\rangle + \mathfrak{F}(v^{k+1}) - \mathfrak{F}(w) \leq 0, \quad \forall w \in H. \quad (4.6)
\]

The theory of [33, 35] provides, under the assumption that \( v^0 \in D(\mathfrak{F}) \) a \( \mathcal{O}(\Delta t^{1/2}) \) error estimate as in Corollary 4.2.

We now introduce the space discretization. Let \( \mathcal{H}_h \) be a family of (finite dimensional) subspaces of \( H \). To be able to handle perturbations on the energy induced by the spatial discretization we assume that, for each \( h > 0 \), we have a convex and lower semicontinuous functional \( \mathfrak{F}_h : \mathcal{H}_h \to \mathbb{R} \) that is \textit{monotone}, in the sense that

\[
\mathfrak{F}_h(w_h) \geq \mathfrak{F}(w_h), \quad \forall w_h \in \mathcal{H}_h. \quad (4.7)
\]

Moreover, we assume that the spaces \( \mathcal{H}_h \) possess suitable approximation properties. In other words, there is a dense subspace \( W \hookrightarrow H \), an operator \( C_h : W \to \mathcal{H}_h \) and functions \( \varepsilon_i \in C((0, \infty), [0, \infty)), \varepsilon_i(0) = 0, i = 1, 2 \), such that

\[
\|C_h w - w\|_H \leq \varepsilon_1(h)\|w\|_W, \quad \forall w \in W, \quad (4.8)
\]

and this approximation is \textit{asymptotically energy diminishing}, i.e.,

\[
\mathfrak{F}_h(C_h w) - \mathfrak{F}(w) \leq \varepsilon_2(h)\|w\|_W, \quad \forall w \in W. \quad (4.9)
\]
We consider the following fully discrete problem: Find \( v_h^{\Delta t} \subset \mathcal{H}_h \) such that
\[
\left< \frac{\delta v_h^{k+1}}{\Delta t}, v_h^{k+1} - w_h \right> + \delta_h(v_h^{k+1}) - \delta_h(w_h) \leq \left< \rho_h^{k+1}, v_h^{k+1} - w_h \right>,
\]
(4.10)
for all \( w_h \in \mathcal{H}_h \), where \( \rho_h^{\Delta t} \subset H \) is a perturbation.

We begin our analysis of the fully discrete method (4.10) with an \textit{a priori} bound on the increments of the sequence \( v_h^{\Delta t} \).

**Lemma 4.4** (Stability of derivatives). The solution \( v_h^{\Delta t} \) to (4.10) satisfies
\[
\frac{1}{2} \| v_h^{\Delta t} \|_{H^{1/2}(H)}^2 + \Delta t \delta_h(v_h^0) \leq \Delta t \delta_h(v_h^0) + \frac{\Delta t}{2} \| \rho^{\Delta t} \|_{L^2(H)}^2.
\]

**Proof.** Set \( w_h = v_h^{k} \) in (4.10). Multiplying by \( \Delta t \), adding over \( k \), and using Young’s inequality, we obtain the result. \( \square \)

The sequence \( \rho^{\Delta t} \) is meant to be a perturbation induced by either discretization or the solution procedure. For this reason we shall assume that
\[
\| \rho^{\Delta t} \|_{L^2(H)} \leq c \Delta t^{1/2}.
\]
(4.11)

Based on this estimate, the error analysis proceeds as follows.

**Theorem 4.5** (A priori error analysis). Let \( v^{\Delta t} \) be the solution to (4.6) and \( v_h^{\Delta t} \) the solution of (4.10). Assume \( v^{\Delta t} \in \ell^\infty(W) \). If the operator \( C_h \) satisfies (4.8) and (4.9); the discrete energies \( \delta_h \) satisfy (4.7); \( \delta_h(v_h^0) < +\infty \) uniformly in \( h \); and the perturbations \( \rho^{\Delta t} \) satisfy (4.11), then there exists a constant \( c > 0 \) proportional to \( T \) such that
\[
\| v^{\Delta t} - v_h^{\Delta t} \|_{L^\infty(H)} \leq \| v^0 - v_h^0 \|_{H}^2 + c(\varepsilon_1(h) + \varepsilon_2(h)) \| v^{\Delta t} \|_{L^\infty(W)} + c \Delta t.
\]

**Proof.** To simplify notation, as usual, we will denote \( e^k = v^k - v_h^k \). Set \( w = v_h^{k+1} \) in (4.6) and \( w_h = C_h v^{k+1} \) in (4.10) and add the results. Using the monotonicity property (4.7) we obtain
\[
\left< \frac{\delta e^{k+1}}{\Delta t}, e^{k+1} \right> \leq \delta_h(C_h v^{k+1}) - \delta_h(v^{k+1}) - \left< \rho^{k+1}, e^{k+1} \right>
\]
\[
+ \left< \rho^{k+1}, v^{k+1} - C_h v^{k+1} \right> + \left< \frac{\delta v^{k+1}}{\Delta t}, C_h v^{k+1} - v^{k+1} \right>.
\]

Using Lemma 4.4 and (4.11) we obtain \( \Delta t^{-1} \sum_{k=1}^K \| \delta v_k \|_{H}^2 \leq c \), whence (4.8) and (4.9) yield
\[
\left< \delta e^{k+1}, e^{k+1} \right> \leq c \Delta t [\varepsilon_2(h) + \varepsilon_1(h)] \| e^{k+1} \|_{W} + \Delta t \| \rho^{k+1} \|_{H} \| e^{k+1} \|_{H}.
\]

Since the sequence \( \| e^{\Delta t} \|_{H} \) is finite we can assume that its maximum is attained for \( k = \kappa \). Summing the above inequality over \( k = 1, \kappa - 1 \) we deduce
\[
\| e^\kappa \|_{H} + \| \delta e^\kappa \|_{H}^2 \leq \| e^0 \|_{H}^2 + c(\varepsilon_1(h) + \varepsilon_2(h)) \| v^{\Delta t} \|_{L^\infty(W)} + 2 \Delta t \| e^\kappa \|_{H} \sum_{k=1}^\kappa \| \rho^k \|_{H}.
\]

An application of Young’s inequality, together with assumption (4.11) allows us to conclude. \( \square \)
4.3. Fully Discrete Scheme for TV Flows. Let us specialize the ideas presented in §4.2 to the case of TV flows. To do so, we will work on the discrete spaces $\mathcal{V}_h$ defined in §2.6, define discrete energies and approximation operators, and verify that assumptions (4.7)–(4.9) are satisfied. The conclusion of Theorem 4.5 will then allow us to obtain error estimates.

In this setting, we consider the subgradient flow: Find $u_h^{\Delta t} \subset \mathcal{V}_h$ that solves

$$
\left\langle \frac{\delta u_h^{k+1}}{\Delta t}, u_h^{k+1} - w_h \right\rangle + \Psi_h(u_h^{k+1}) - \Psi_h(w_h) \leq 0, \quad \forall w_h \in \mathcal{V}_h.
$$

(4.12)

The discrete energy is defined by

$$
\Psi_h(w_h) = \sum_{T \in \mathcal{T}_h} \int_T I_h |\nabla w_h|,
$$

(4.13)

where the operator $I_h$ was defined in (2.7). Notice that we have effectively replaced the total variation seminorm by a quadrature formula. Indeed, we can rewrite $\Psi_h$ as

$$
\Psi_h(w_h) = \sum_{T \in \mathcal{T}_h} \sum_{j=1}^{N_T} |\nabla w_h|(z_{j,T}) \int_T \lambda_{j,T}(z) \, dz = \sum_{T \in \mathcal{T}_h} |T| \sum_{j=1}^{N_T} \omega_j |\nabla w_h|(z_{j,T}),
$$

where $\lambda_{j,T}$ is the coordinate basis function associated with node $z_{j,T}$ and we denote the weights by $\omega_j = |T|^{-1} \int_T \lambda_{j,T}(z) \, dz$. Notice that this modification fits into the framework described in (4.12). Its utility will become clear in the following paragraph.

The approximation operator is defined as

$$
C_h w = \Pi_h w_c,
$$

(4.14)

where $\Pi_h$ is the Clément interpolation operator and $w_c$ denotes a regularization of $w$ as in Proposition 2.1 with $\epsilon = h^{2/3}$. Proving error estimates reduces to verification of the hypotheses of §4.2.

Corollary 4.6 (Convergence of TV flow). Let $u^{\Delta t}$ be the solution of (4.1) and $u_h^{\Delta t}$ the solution of (4.12) with the discrete energy given in (4.13). If $\Omega$ is star shaped with respect to a point; $u^0 \in L^\infty(\Omega) \cap BV(\Omega)$; and $\Psi_h(u_h^0) \leq c < +\infty$ uniformly in $h$, then

$$
\| u^{\Delta t} - u_h^{\Delta t} \|_{L^\infty(L^2)}^2 \leq \| u^0 - u_h^0 \|_{L^2}^2 + ch^{1/3}.
$$

Proof. Notice, first of all, that the given assumptions on $u^0$ translate into $u^{\Delta t} \in L^\infty(L^\infty(\Omega) \cap BV(\Omega))$. Consequently, we set $W = BV(\Omega) \cap L^\infty(\Omega)$ and

$$
\| w \|_W = \| w \|_{L^\infty} + \| D w \|_{L^\infty(\Omega)}.
$$

It suffices to verify the abstract assumptions (4.7)–(4.9):

1. If $w_h |_T \in P_1$, then $\nabla w_h$ is constant and $I_h |\nabla w_h| = |\nabla w_h|$. If $w_h |_T \in Q_1$ instead, then its gradient is linear and, consequently, $|\nabla w_h|$ is convex. Then, we only need to realize that if the function $\phi$ is convex, then $I_h \phi |_T \geq \phi |_T$.

Indeed, using that $\sum_{j=1}^{N_T} \lambda_{j,T} \equiv 1$,

$$
I_h \phi |_T(z) = \sum_{j=1}^{N_T} \phi(z_{j,T}) \lambda_{j,T}(z) \geq \phi \left( \sum_{j=1}^{N_T} z_{j,T} \lambda_{j,T}(z) \right) = \phi(z).
$$
\[ L^8 \colon (2.1) \text{, together with the } L^\infty \text{-stability of } \Pi_h, \text{ gives} \]
\[ \| \Pi_h w_\epsilon - w \|_{L^2} \leq \| \Pi_h w_\epsilon - w \|_{L^1} \| \Pi_h w_\epsilon - w \|_{L^\infty} \leq c \| w \|_{L^\infty} \| \Pi_h w_\epsilon - w \|_{L^1}. \]

To bound the \( L^1 \) norm in the above inequality we add and subtract \( w_\epsilon \) to obtain
\[ \| \Pi_h w_\epsilon - w \|_{L^1} \leq \| \Pi_h w_\epsilon - w_\epsilon \|_{L^1} + \| w_\epsilon - w \|_{L^1} \leq c \left( \frac{h^2}{\epsilon} + \epsilon \right) \| Dw \| (\Omega), \]

where we have used \( (2.8), \) for \( p = 1, \) in conjunction with Proposition \( 2.1 \)

\[ L^9 \colon \text{We begin the proof of the energy diminishing property by} \]
\[ \Psi_h(\Pi_h w_\epsilon) - \Psi(w) = \int_\Omega I_h |\nabla \Pi_h w_\epsilon| - |Dw| (\Omega) \leq \]
\[ \left( \int_\Omega (I_h |\nabla \Pi_h w_\epsilon| - |\nabla \Pi_h w_\epsilon|) \right) + \left( \int_\Omega (|\nabla \Pi_h w_\epsilon| - |\nabla w_\epsilon|) \right) + \]
\[ \left( \int_\Omega |\nabla w_\epsilon| - |Dw| (\Omega) \right) = I + II + III. \]

Applying Proposition \( 2.1 \) yields \( III \leq c \| Dw \| (\Omega). \) We next invoke the triangle inequality along with interpolation estimate \( (2.8) \) for \( \Pi_h \) to arrive at
\[ II \leq \int_\Omega |\nabla (\Pi_h w_\epsilon - w_\epsilon)| \leq c h / \epsilon \| D^2 w_\epsilon \|_{L^1} \leq c h / \epsilon \| Dw \| (\Omega). \]

For the first term \( I, \) we use \( (2.8) \) to obtain
\[ I = \int_\Omega I_h |\nabla \Pi_h w_\epsilon| - |\nabla \Pi_h w_\epsilon| \leq c h \sum_{T \in \mathcal{T}_h} \| \nabla (|\nabla \Pi_h w_\epsilon|) \|_{L^1(T)}. \]

For a smooth function,
\[ \nabla (|\nabla f|) = \frac{\nabla f}{|\nabla f|} D^2 f, \quad \text{a.e. } \{ x \in T : \nabla f(x) \neq 0 \}, \]
and \( \nabla (|\nabla f|) = 0 \text{ a.e. } \{ x \in T : \nabla f = 0 \}. \) This allows us to conclude that \( |\nabla (|\nabla f|)| \leq |D^2 f| \text{ a.e. in } T. \) This, together with the bound \( (2.9) \) and Proposition \( 2.1 \) shows that
\[ \int_\Omega I_h |\nabla \Pi_h w_\epsilon| - |\nabla \Pi_h w_\epsilon| \leq c h |D^2 w_\epsilon|_{L^1} \leq c h / \epsilon |Dw| (\Omega). \]

The estimates above allow us to see that
\[ \epsilon_1(h) = c \left( \frac{h^2}{\epsilon} + \epsilon \right)^{1/2} \quad \epsilon_2(h) = c \left( \frac{h}{\epsilon} + \epsilon \right). \]

Setting \( \epsilon = h^{2/3} \) we obtain the result. \( \square \)

**Remark 4.7 (Energy diminishing interpolation).** The proof of Corollary \( 4.6 \) actually shows that if we were able to construct a TV diminishing interpolant, i.e., such that
\[ \int_\Omega |\nabla C_h w| \leq |Dw| (\Omega), \]
then \( \epsilon_2(h) = c h^{-1} \) and, setting \( \epsilon = h^{1/2} \) we would improve the rate of convergence to \( O(h^{1/4}) \). Under the assumption that \( \Omega = (0,1)^d \) and that the mesh is Cartesian, \( (32) \) presents such a construction.
5. An Iterative Scheme for the Fully Discrete Scheme. A practical algorithm for the solution of the discrete variational inequalities \((4.12)\) would require identifying the elements of the subdifferential \(\partial \Psi_h(u)\). In the continuous setting, relying on the results of [3], such identification is presented in [2]. Let us describe this identification without going into technical details. We introduce the space

\[ X = \{ q \in L^\infty(\Omega) : \nabla q \in L^2(\Omega), \; q \cdot n = 0 \}, \]

and stress that \(w \in \partial \Psi(u)\) if and only if [2]

\[ \Psi(u) = \int_{\Omega} w u, \quad \text{and} \quad \exists z \in X \cap \mathcal{B}_1(L^\infty(\Omega)) : \; w = -\nabla z. \]

Notice that we can replace the first equality above by \(\Psi(u) = \int_{\Omega} z : Du\) [2]. This serves as motivation for the discrete energy \(\Psi_h\) given in (4.13). Indeed,

\[ \Psi_h(w_h) = \sum_{T \in T_h} |T| \sum_{j=1}^{N_T} \omega_j \nabla w_h(z_j) : (z_j) = \sum_{T \in T_h} |T| \sum_{j=1}^{N_T} \omega_j \sup_{q \in \mathbb{R}^d, |q| \leq 1} \nabla w_h(z_j) \cdot q \]

\[ = \sup_{q_h \in X_h : \|q_h\|_{X_h} \leq 1} \langle \nabla w_h, q_h \rangle_h = \sup_{q_h \in X_h} \{ \langle \nabla w_h, q_h \rangle_h - \mathcal{J}_S(q_h) \}, \]

where \(\mathcal{J}_S\) denotes the indicator function of \(S\), the discrete space \(X_h\) is defined as

\[ X_h = \{ q_h \in L^\infty(\Omega) : q_h|_T \in \mathbb{Q}_1^d, \forall T \in T_h \}, \quad \|q_h\|_{X_h} = \max_{T \in T_h} \max_{j=1}^{N_T} \{ |q_h(z_j)| \}, \]

and the discrete inner product \(\langle \cdot, \cdot \rangle_h\) is defined by the quadrature rule

\[ \langle q_h, r_h \rangle_h = \sum_{T \in T_h} |T| \sum_{j=1}^{N_T} \omega_j q_h(z_j) \cdot r_h(z_j). \]

The latter induces the norm \(\|q_h\|_h^2 = \langle q_h, q_h \rangle_h\), which clearly implies

\[ \|q_h\|_h \leq c \|q_h\|_{L^2}, \quad \forall q_h \in X_h. \quad (5.1) \]

In this setting it is not difficult to see that the fully discrete subgradient flow \((4.12)\) with energy \((4.13)\) is equivalent to finding \(u_h^{k+1} \in \mathbb{V}_h\) and \(z_h^{k+1} \in \mathcal{B}_1(X_h)\) that solve:

\[ \left\langle \frac{\delta u_h^{k+1}}{\Delta t}, w_h \right\rangle + \langle z_h^{k+1}, \nabla w_h \rangle_h = 0, \quad \forall w_h \in \mathbb{V}_h, \quad (5.2) \]

and

\[ \langle q_h - z_h^{k+1}, \nabla u_h^{k+1} \rangle_h \leq 0, \quad \forall q_h \in \mathcal{B}_1(X_h). \quad (5.3) \]

5.1. Exact Solver. Problem \((5.2) - (5.3)\) is not a practical numerical scheme, since it involves the solution of the (local) variational inequality \((5.3)\). To overcome this difficulty we exploit that \((5.2) - (5.3)\) are the optimality conditions of the functional

\[ w_h \mapsto \frac{1}{2N} \|w_h - u_h^0\|_{L^2}^2 + \int_{t_0}^{t_h} \mathcal{T}_h |\nabla w_h|, \]

which motivates the following algorithm for the
solution of (5.2–5.3) (cf. [7, 12]): Let \( \sigma > 0 \) and \( \tau > 0 \) be parameters to be chosen. Given \( g_h = u_h^0 \in V_h \), set \( v_h^0 = g_h \), \( \lambda_h^0 = 0 \) and find \( \{v_h^l, \lambda_h^l\} \subset V_h \times B_1(X_h) \) by

\[
\left\langle -\frac{\sigma}{\tau} \delta \lambda_h^{l+1} + \nabla v_h^{l+1}, \mathbf{q}_h - \lambda_h^{l+1} \right\rangle_h \leq 0 \quad \forall \mathbf{q}_h \in B_1(X_h),
\]

where \( v_h^{l+1} \) is the extrapolation defined in (2.3), and

\[
\left\langle \frac{\delta v_h^{l+1}}{\tau}, w_h \right\rangle + \left\langle \lambda_h^{l+1}, \nabla w_h \right\rangle_h + \frac{1}{\Delta t} \left\langle v_h^{l+1}, w_h \right\rangle = \frac{1}{\Delta t} \left\langle g_h, w_h \right\rangle.
\]

Finally, set \( u_h^{k+1} = v_h^\infty \).

The last assertion needs to be justified, this is given in the following.

**Theorem 5.1** (Convergence of the inner iterations). *Assume the inverse inequality \( \|\nabla w_h\|_{L^2} \leq c_h^{-1} \|w_h\|_{L^2} \) holds, and the parameters \( \tau \) and \( \sigma \) are chosen so that \( \tau \leq c_1 \sigma h \). Then the inner iterative loop (5.4)–(5.5) converges, in the sense that for every \( L \geq 1 \),

\[
\sum_{l=0}^{L} \left( \frac{\tau}{\Delta t} \|u_h^{l+1} - v_h^{l+1}\|_{L^2}^2 + \frac{\sigma}{2} \|\delta \lambda_h^{l+1}\|_{h}^2 \right) \leq c.
\]

**Proof.** We follow the arguments of [7, Proposition 3.1]. To alleviate the notation, set \( e_h^l = u_h^{k+1} - v_h^{l} \) and \( E_h^l = z_h^{k+1} - \lambda_h^{l} \). Subtract (5.5) from (5.2) to obtain

\[
\left\langle \frac{\delta e_h^{l+1}}{\tau}, w_h \right\rangle + \left\langle E_h^{l+1}, \nabla w_h \right\rangle_h + \frac{1}{\Delta t} \left\langle e_h^{l+1}, w_h \right\rangle = 0.
\]

Set \( \mathbf{q}_h = z_h^{k+1} \) in (5.4), \( \mathbf{q}_h = \lambda_h^{l+1} \) in (5.3) and add them, to obtain

\[
- \left\langle -\frac{\sigma}{\tau} \delta \mathbf{E}_h^{l+1} + \nabla \left( u_h^{k+1} - v_h^{l+1} \right), \mathbf{E}_h^{l+1} \right\rangle_h \leq 0.
\]

Set \( w_h = 2\tau e_h^{l+1} \) in (5.6) to obtain

\[
\|e_h^{l+1}\|_{L^2}^2 + \|\delta e_h^{l+1}\|_{L^2}^2 + 2\tau \left( \|E_h^{l+1} + \nabla e_h^{l+1}\|_h^2 + \frac{2\tau}{\Delta t} \|e_h^{l+1}\|_{L^2}^2 \right) \leq 0.
\]

Multiply (5.7) by \( 2\tau \) and add it to (5.8) and add over \( l = 0, L - 1 \) to obtain

\[
\|e_h^L\|_{L^2}^2 + \sigma \|E_h^L\|_{h}^2 + \sum_{l=0}^{L-1} \left( \|e_h^{l+1}\|_{L^2}^2 + \sigma \|\delta E_h^{l+1}\|_{h}^2 + \frac{2\tau}{\Delta t} \|e_h^{l+1}\|_{L^2}^2 \right) \leq \|e_h^0\|_{L^2}^2 + \sigma \|E_h^0\|_{h}^2 + 2\tau \sum_{l=0}^{L-1} \left\langle \delta^2 \nabla e_h^{l+1}, E_h^{l+1} \right\rangle_h.
\]

Now we proceed as in [7, Proposition 3.1] with the last term: we use (2.5) to sum by parts and next employ inequality (5.1), the inverse inequality and repeated applications of Cauchy-Schwarz to obtain the desired result. \( \square \)

**Remark 5.2** (Convergence of the Lagrange multiplier). *As already mentioned in [7, Remark 3.2(i)], convergence \( \lambda_h \to z_h^{k+1} \), as \( l \to \infty \), cannot be expected in general,
due to the non-uniqueness of $\tilde{z}_h^{k+1}$. However the increments $\delta \lambda_h^k$ converge to zero.

**Remark 5.3 (Solution of the discrete variational inequality).** It is not difficult to see that the solution to the discrete variational inequality (5.4) is explicit

$$
\lambda_h^{i+1}(z_j, t) = \frac{\lambda_h^i + \frac{\sigma}{\tau} \nabla v_h^{r,i+1}}{\max\{1, |\lambda_h^i + \frac{\sigma}{\tau} \nabla v_h^{r,i+1}|(z_j, t)\}}.
$$

### 5.2. Inexact Solver

It is not feasible to iterate in (5.4)–(5.5) until convergence. For this reason we included in the analysis presented in §4.2 a perturbation term since if we were to stop after $L - 1$ iterations, (5.5) would become

$$
\frac{1}{\Delta t} \left( \langle v_h^k - u_h^k, w_h \rangle + \langle \lambda_h^k, \nabla w_h \rangle \right)_h = -\frac{1}{\tau} \langle \delta v_h^k, w_h \rangle,
$$

and (5.4) would read

$$
\left( r_h^k + \nabla v_h^k, q_h - \lambda_h^k \right)_h \leq 0,
$$

with $r_h^k = -\frac{\sigma}{\tau} \delta \lambda_h^k - \delta^2 \nabla v_h^k$. Doing the formal replacements $v_h^k \leftarrow \tilde{u}_h^{k+1}$ and $\lambda_h^k \leftarrow \tilde{z}_h^{k+1}$ on the left hand side of these identities, setting $w_h = \tilde{u}_h^{k+1} - \tilde{w}_h$ and adding them we obtain

$$
\left( \frac{\delta \tilde{u}_h^{k+1}}{\Delta t}, \tilde{u}_h^{k+1} - \tilde{w}_h \right) + \tilde{\Psi}_h(\tilde{u}_h^{k+1}) - \tilde{\Psi}_h(\tilde{w}_h) \leq -\left( \frac{\delta v_h^k}{\tau}, \tilde{u}_h^{k+1} - \tilde{w}_h \right),
$$

(5.9)

with

$$
\tilde{\Psi}_h(w_h) = \sum_{T \in T_h} \int_T I_h|\nabla w_h + r_h^k|.
$$

The non-homogeneous term on the right hand side can be understood as a perturbation $\rho^{\Delta t}$. In addition, the modified discrete energies can be controlled owing to Theorem 5.1. We make these ideas rigorous in the following.

**Theorem 5.4 (Convergence of fully discrete scheme with inexact solutions).** Let $\Omega$ be star-shaped with respect to a point and assume that $u_0 \in BV(\Omega) \cap L^\infty(\Omega)$. If $\{\tilde{u}_h^{\Delta t}, \tilde{z}_h^{\Delta t}\} \subset V_h \times X_h$ are approximations to the solution of (5.2)–(5.3), computed with algorithm (5.4)–(5.5) in such a way that, for every time step $k$, the inequalities

$$
\|\tau^{-1} \delta v_h^k\|_{L^2} \leq c\Delta t^{1/2}, \quad \|r_h^k\|_{L^1} \leq c\Delta t,
$$

(5.10)

are satisfied, then the following error estimate holds

$$
\|u - \tilde{u}_h^{k+1}\|_{L^\infty(L^2)} \leq \|u_0 - u_h^0\|_{L^2} + c(\Delta t^{1/2} + h^{1/6}).
$$

The spatial rate of convergence becomes $O(h^{1/4})$ as soon as $\Omega = (0, 1)^d$ and the mesh $T_h$ is Cartesian.

**Proof.** Since $u_0 \in BV(\Omega) \cap L^\infty(\Omega)$, the solution, $u^{\Delta t}$, to (4.1) converges, with order $O(\Delta t^{1/2})$, to the exact solution. This also guarantees, via Corollary 4.6, the convergence of $u_h^{\Delta t}$, solution of (4.12), to $u^{\Delta t}$ with order $O(h^{1/6})$. 


To study the convergence of $\tilde{u}^k$ to $u^k$ we proceed as in Theorem 4.5. Set $w_h = \tilde{u}^{k+1}$ in (4.12) and $\tilde{w}_h = u^{k+1}$ in (5.9) and add the result. On denoting $e^k = u^k - \tilde{u}_h^k$, we obtain

$$
\frac{1}{\Delta t} \langle \delta e^k, e^k \rangle + \Psi_h(u_h^{k+1}) - \tilde{\Psi}_h(u^{k+1}) \leq \left\langle \frac{\delta u^L}{\tau}, e^k \right\rangle.
$$

Since

$$
\tilde{\Psi}_h(w_h) - \Psi_h(w_h) = \sum_{T \in T_h} \int_T (\mathcal{I}_h \nabla w_h + r_h^T) - \mathcal{I}_h \nabla w_h) \leq c \|r_h^T\|_{L^1},
$$

we notice that condition (5.10) suffices to guarantee that $\|e^k\|_{L^\infty(L^2)} \leq c \Delta t^{1/2}$. Conclude with a trivial application of the triangle inequality.

The improvement on the spatial rate of convergence is due to Remark 4.7.

6. Numerical Experiments. To illustrate the theory developed in the preceding sections, here we present a series of numerical experiments. We implemented scheme (5.4)–(5.5) where the stopping criterion for the inner iterations is given by (5.10). The implementation was done with the help of the deal.2 library (see §4, §5). Unless noted otherwise, we set $\Delta t = \frac{\sqrt{2h}}{m}$, $\tau = \Delta t$ and $\sigma = 0.1$.

6.1. The Characteristic of a Convex Set. Let $E \subset \Omega \subset \mathbb{R}^2$, where $E$ is convex, connected and of finite perimeter. Define $\lambda_E = P(E)/|E|$, where $P(E)$ stands for the perimeter of $E$. Assume, in addition, that $\partial E \in C^{1,1}$ and that the curvature of $E, \kappa$, satisfies $\|\kappa\|_{L^\infty(\partial E)} \leq \lambda_E$. Then, according to [2, 3], if $u_0 = \chi_E$ the solution to (3.2) with homogeneous Dirichlet boundary conditions is $u(x,t) = (1 - \lambda_E t)^+ \chi_E(x)$.

Figure 6.1 shows the $L^\infty(L^2)$-error between the discrete and numerical solutions in the case when $\Omega = (-3,3)^2$ and $E = B(0,1)$. As we can see the behavior of the error is actually better (i.e., $O(h^{1/2})$) than what our theory predicts.

6.2. The Characteristic of a Disconnected Set. With the setting of 6.1, references [2, 3] also show that if $u_0 = \sum_{i=1}^n \chi_{B(x_i, r_i)}$, where the centers are at the vertices of an equilateral triangle of unit side-length and the radii satisfy $r_i \leq 0.2$, then $u(x,t) = \sum_{i=1}^3 (1 - \lambda_B t)^+ \chi_{B(x_i, r_i)}$.

Fig. 6.1. $L^\infty(L^2)$-errors for the evolution of the characteristic $\mathcal{X}_E$ of a circle $B$ with exact solution $u(x,t) = (1 - \lambda_E t)^+ \chi_E(x)$ and $\lambda_E = P(B)/|B|$ (see §4, §5).

Figure 6.1 shows the $L^\infty(L^2)$-error between the discrete and numerical solutions in the case when $\Omega = (-3,3)^2$ and $E = B(0,1)$. As we can see the behavior of the error is actually better (i.e., $O(h^{1/2})$) than what our theory predicts.
Fig. 6.2. Evolution of the characteristic set of three balls \( B_i = B(x_i, r_i) \) with centers \( x_i \) at the vertices of an equilateral triangle of unit size and radii \( r_i \leq 0.2 \). The exact solution is
\[
u(x,t) = \sum_{i=1}^{3} (1 - \lambda_{B_i}) t^i \chi_{B(x_i, r_i)}.
\]
The mesh is uniform with size \( h = 2^{-5} \) and the time-step \( \Delta t = \sqrt{2} h / 10 \). The discrete solution is shown every three time steps until extinction and preserves the structure of \( u \) without numerical diffusion and/or oscillations.

Figure 6.2 shows the evolution in this case when \( h = 0.03 \). Notice that our method provides a good approximation with relatively few nodes and it does not introduce spurious oscillations. In addition, we obtain fairly good agreement with the extinction time, given by \( t_* = \frac{1}{\lambda_B} = \frac{r}{2} = \frac{1}{10} \). In contrast to methods that involve regularization \([21, 22]\), we observe consistency with the PDE in that the support of the fully discrete solution remains constant in time. Finally, the \( L^\infty(L^2) \)-norm of the error with respect to \( h \) is shown in Figure 6.3. Again, we see that the error is \( O(h^{1/2}) \).

6.3. The Characteristic of an Annulus. Again in the setting of \([6.1]\), the evolution of \( u_0 = M \chi_E \), where \( E = B(0, R) \setminus B(0, r) \) is an annulus and \( M, R, r \in \mathbb{R} \),
r < R, is governed by \[ u(x,t) = \begin{cases} 
\text{sgn}(M)(|M| - \lambda_E t)^+ \chi_E + \lambda_B(t \chi_B), & t < T_1, \\
\text{sgn}(m) \left(|m| - \lambda_B(t - T_1)^+ \chi_B \right), & t \geq T_1,
\end{cases} \]
with \( T_1 = \frac{|M|}{\lambda_E + \lambda_B(0,r)} \) and \( m = \lambda_B(0,r) T_1 \).

We set \( M = 4, R = 1/2, r = 1/4 \). Figure 6.4 shows the evolution of the numerical solution for \( h = 2^{-5} \) and notice that, in this case, \( T_1 = 1/4, m = 2 \) and the extinction time is \( T_{ext} = 3/4 \). Numerically we obtained \( T_{1,h} \approx 0.2558, m_h \approx 2.045, T_{ext,h} \approx 0.796 \); which are in good agreement with the exact values. In addition, we see that there is no diffusion, as opposed to what methods based on regularization obtain. The \( L^\infty(L^2) \)-error with respect to \( h \) is presented in Figure 6.5. Again we obtain \( O(h^{1/2}) \).

### 6.4. Convergence of the Inner Loop

The result of Theorem 5.1 guarantees convergence of the inner iteration at every step. However, there is no assessment of the speed of convergence. We numerically investigate the effect of the choice of initial condition for the inner loop. Set \( \Omega = (-2,2) \) and \( u_0 = \chi_{[-1,1]} \). We choose
\[
\begin{align*}
\mathbf{z}_0^1 &= 0, \\
\mathbf{z}_0^2 &= \begin{cases} 
\mathbf{x} + 2, & \mathbf{x} \in (-2,-1), \\
-\mathbf{x}, & \mathbf{x} \in (-1,1), \\
\mathbf{x} - 2, & \mathbf{x} \in (1,2)
\end{cases}.
\end{align*}
\]

It is possible to show that only \( \mathbf{z}_0^2 \in \partial \Psi(u_0) \). For \( h = 2^{-6}, \Delta t = \tau = h/10 \) and \( \sigma = 1 \), Figure 6.6 plots the number of iterations as a function of the step for \( \mathbf{z}_0^1 \) and \( \mathbf{z}_0^2 \). The number of initial iteration is much higher (more than 5000) for \( \mathbf{z}_0^1 \) than for \( \mathbf{z}_0^2 \). Moreover, notice the spike on both graphs at 79 steps. This is due to the fact that, at this step, extinction occurs. From this we conclude that the number of iterations heavily depends on the initial choice of \( \mathbf{z}_0 \) or, more generally, on \( \lambda_0^0 \). Strategies for choosing it need further investigation.

### 6.5. Comparison with Regularized Flow

To conclude our discussion, it is imperative to make a comparison between our method and those that involve regularization and show that the advantages in our approach are numerous. Before embarking in such an endeavor, let us recall some properties of regularized flows. The analysis developed in [22] does not provide a clear understanding of the relation between the
regularization $\epsilon$ and the space discretization $h$. By means of numerical experiments the authors conclude that $\epsilon = O(h^2)$ is the optimal scaling law, see [22, Figures 4–5].

Let us, with the help of the theory developed in §4.2, try to bring some light into this matter. If we denote by $u_\epsilon$ the solution to (1.1) with regularization $\Psi_\epsilon(w) = \int_\Omega \sqrt{\epsilon^2 + |\nabla w|^2}$ given by (1.2), [22, Theorem 2] shows that

$$\|u - u_\epsilon\|_{L^\infty(L^2)} \leq c\epsilon^{1/2}.$$  \hspace{1cm} (6.1)

If $u_{\epsilon,h}^0 \subset V_h$ is the solution to a fully discrete approximation of the regularized flow with $\Psi_{\epsilon,h} = \Psi_\epsilon$, under the assumption that $u_\epsilon \in L^\infty(0,T;H^2(\Omega))$ we have

$$\|u_\epsilon - u_{\epsilon,h}^0\|_{L^\infty(L^2)} \leq c \left( \Delta t^{1/2} + h^{1/2}\epsilon^{-\alpha/2} \right).$$  \hspace{1cm} (6.2)

To see this it suffices to realize that (4.7) is trivially satisfied and, setting $C_h = \Pi_h$ we conclude that (4.8) amounts to

$$\|I_h w - w\|_{L^2} \leq c h^2 \|w\|_{H^2} \quad \Rightarrow \quad \varepsilon(h) = c h^2 \|w\|_{H^2},$$

and, since

$$\int_\Omega \sqrt{\epsilon^2 + |\nabla w_1|^2} - \sqrt{\epsilon^2 + |\nabla w_2|^2} \leq \int_\Omega |\nabla (w_1 - w_2)|,$$
we have, for (4.9),
\[ \varepsilon_2(h) \leq ch \| w \|_{W^1_2}. \]

Finally, [21, Theorem 1.2] shows that \( \| u_c \|_{L^2(H^2)} \leq \varepsilon^{-\alpha} \) for some \( \alpha \in \mathbb{N}_0 \). Combining (6.1) and (6.2) with Theorem 4.5 we conclude
\[ \| u - \hat{u}_c \|_{L^\infty(L^2)} \leq c \left( \Delta t^{1/2} + \varepsilon^{1/2} + h \varepsilon^{-\alpha/2} + h^{1/2} \right), \quad (6.3) \]

which yields the optimal scaling \( \varepsilon = O(h^{\frac{1}{1+\alpha}}) \).

The last ingredient we need to make the comparison is to recall that (cf. [28, Remark 5.3]) in the one dimensional case, i.e., \( d = 1 \), if the initial data is monotone, it is itself a minimizer of the total variation energy, and so the flow fixes it. In other words, if \( u_0 \) is monotone, then \( u(t) = u_0 \) for all \( t > 0 \).

Consider, in \( \Omega = (0,1) \), the initial data
\[ u_0(x) = \begin{cases} 
0, & x \leq \frac{1}{2}, \\
 x - \frac{1}{2}, & \frac{1}{2} < x \leq 1.
\end{cases} \]

According to the discussion presented above, the solution to (3.2) is \( u(t) = u_0 \). Figure 6.7 shows the solution, at \( T = 5 \), obtained with our method and the regularized
flow with $\epsilon = O(h^2)$ and $\epsilon = O(h)$. The first choice is the one advocated in [22], while the second is the optimal according to (6.3) provided $\alpha = 1$, which is consistent with Proposition 2.1. In such a case, (6.3) gives the following error estimate

$$\|u - \hat{u}_{\epsilon,h}\|_{L^\infty(L^2)} \leq ch^{1/2},$$

provided $\epsilon \approx h \approx \Delta t$. The mesh size is $h = 2^{-5}$ and the time-step $\Delta t = 2^{-10}$. Notice that the requirement $\Delta t = O(h^2)$, which is needed for $L^2$-convergence in the regularized flow (cf. [22, Theorem 4] and [21, Theorem 1.7]), is satisfied.

The advantages of our method are now evident. We do not impose any restriction on the time-step, as opposed to the $\Delta t = O(h^2)$ that is necessary in regularized methods to guarantee convergence in $L^2(\Omega)$. Even if one is willing to settle for convergence in $L^p(\Omega)$, with $p < d/(d-1)$, the methods with regularization require that the solution of the regularized flow is in $L^\infty(0,T;W^1_1(\Omega)) \cap L^\infty(0,T;H^{1,\text{loc}}(\Omega))$. If an error estimate is desired, one must impose that $u_0 \in C^2(\bar{\Omega})$, and even in that case it is not clear what is the relation between $h$, $\Delta t$ and $\epsilon$. In addition to these approximation issues, the regularized flow requires the solution, at each time step, of a nonlinear system and no convergence analysis is provided. In contrast, we have developed and analyzed an inexact iterative scheme for the solution of our problems at each time step, and we have showed its global convergence. To finalize, the result presented in Figure 6.7 shows that the regularized flow misses certain fundamental features of the problem, even in simple cases.

7. Total Variation Minimization. We conclude with yet another application of our result on approximation of functions of bounded variation (Proposition 2.1): we improve on the existing results about total variation minimization. Let $g \in L^\infty(\Omega)$ and $\alpha > 0$. Consider $\Xi(w) = \Psi(w) + \frac{\alpha}{2}\|w - g\|_{L^2}^2$. Thanks to the fact that this functional is strictly convex there exists a unique $\xi \in BV(\Omega) \cap L^2(\Omega)$ such that

$$\Xi(\xi) = \inf \left\{ \Xi(w) : w \in L^2(\Omega) \right\} < \infty. \quad (7.1)$$

Here we are interested in the approximation of $\xi$ by elements of $V_h$. Since $V_h$ is finite dimensional, there is a unique $\xi_h \in V_h$ such that

$$\Xi(\xi_h) = \inf \left\{ \Xi(w_h) : w \in V_h \right\} < \infty. \quad (7.2)$$
The main approximation properties of $\xi_h$ are detailed in the following.

**Theorem 7.1 (Convergence of discrete minimizers).** Assume that $\Omega$ is star shaped with respect to a point. Let $\xi \in BV(\Omega) \cap L^\infty(\Omega)$ and $\xi_h \in V_h$ be defined as in (7.1) and (7.2), respectively. Then

$$||\xi - \xi_h||_{L^2} \leq ch^{1/4}.$$  

**Proof.** We adapt the ideas presented in [7, Theorem 3.1]. Let $\epsilon > 0$ and $\xi_\epsilon \in C^\infty(\Omega)$ be an approximation of $\xi$ that satisfies all the properties stated in Proposition 2.1. Owing to the strict convexity of $\Xi$ and the fact that $\xi_h$ is a discrete minimizer, we have

$$\alpha \frac{1}{2} ||\xi - \xi_h||^2_{L^2} \leq \Xi(\xi_h) - \Xi(\xi) = (||\nabla \Pi_h \xi_\epsilon||_{L^1} - |D\xi|(\Omega)) + \alpha \frac{1}{2} (||\Pi_h \xi_\epsilon - g||^2_{L^2} - ||\xi - g||^2_{L^2}) = A_1 + A_2,$$

where $\Pi_h$ is the Clément interpolation operator [15]. Let us look at each one of the terms in this last inequality:

$A_1$: We add and subtract the $W^{1,1}$-seminorm of $\xi_\epsilon$ and use its approximation properties with respect to total variation along with the approximation properties of $\Pi_h$ described in (2.8):

$$A_1 \leq ||\nabla (\Pi_h \xi_\epsilon - \xi_\epsilon)||_{L^1} + ||\nabla \xi_\epsilon||_{L^1} - |D\xi|(\Omega) \leq c \left( \frac{h}{\epsilon} + \epsilon \right) |D\xi|(\Omega).$$

$A_2$: We, again, use the approximation properties of $\xi_\epsilon$ and the fact that $\xi$ and $g$ are essentially bounded, say by a constant $c > 0$,

$$A_2 \leq \frac{c\alpha}{2} (||\nabla \xi_\epsilon - \xi_\epsilon||_{L^1} + ||\xi_\epsilon - \xi||_{L^1}) \leq c \left( \frac{h^2}{\epsilon} + \epsilon \right) |D\xi|(\Omega).$$

Setting $\epsilon = h^{1/2}$ we obtain the result. □

**Remark 7.2 (Convergence of total variation minimization).** Theorem 7.1 is, in a sense, an improvement over the original result of [7], at least for star shaped domains, and under the boundedness assumptions on $g$ and $\xi$. If $\xi \in B^{s_\infty}_s(L^2(\Omega))$ for some $s \in (0, 1]$, and relying on the results of [32], [7, Theorem 3.1] proves the estimate

$$||\xi - \xi_h||_{L^2} \leq ch^{1/(2s+1)},$$

so that the best possible rate of convergence is $O(h^{1/4})$, which is what we obtain, but with lower regularity. To understand this regularity assumption it suffices to recall that $BV(\Omega) \cap L^\infty(\Omega) \hookrightarrow B^{s_\infty}_s(L^2(\Omega))$ for $s \leq \frac{1}{2}$ (see [32], Lemma 3.8.1] for a proof and, in some sense, the converse inclusion). The key step that allowed us to reduce the smoothness assumption is Proposition 2.1. In addition, the proof of Theorem 7.1 shows that if we had a TV-diminishing interpolant, we would obtain

$$||\xi - \xi_h||_{L^2} \leq ch^{1/2},$$

which is an optimal error estimate for $\xi \in B^{s_\infty}_s(L^2(\Omega))$. Such a construction is presented in [32].
REFERENCES

[1] L. Ambrosio, N. Fusco, and D. Pallara, *Functions of bounded variation and free discontinuity problems*, Oxford Mathematical Monographs, The Clarendon Press Oxford University Press, New York, 2000.

[2] F. Andreu-Vaillo, V. Caselles, and J.M. Mazón, *Parabolic quasilinear equations minimizing linear growth functionals*, vol. 223 of Progress in Mathematics, Birkhäuser Verlag, Basel, 2004.

[3] G. Anzellotti, *Pairings between measures and bounded functions and compensated compactness*, Ann. Mat. Pura Appl. (4), 135 (1983), pp. 293–318 (1984).

[4] W. Bangerth, R. Hartmann, and G. Kanschat, *deal.II Differential Equations Analysis Library*, Technical Reference. http://www.dealii.org.

[5] ———, *deal.II — a general-purpose object-oriented finite element library*, ACM Trans. Math. Softw., 33 (2007).

[6] V. Barbu, *Nonlinear differential equations of monotone types in Banach spaces*, Springer Monographs in Mathematics, Springer, New York, 2010.

[7] S. Bartels, *Total variation minimization with finite elements: Convergence and iterative solution*, SIAM Journal on Numerical Analysis, 50 (2012), pp. 1162–1180.

[8] G. Bellettini, V. Caselles, and M. Novaga, *The total variation flow in $\mathbb{R}^N$*, J. Differential Equations, 184 (2002), pp. 475–525.

[9] M. Breuss, T. Brox, A. Burgel, T. Sonar, and J. Weickert, *Numerical aspects of TV flow*, Numer. Algorithms, 41 (2006), pp. 79–101.

[10] H. Brézis, *Opérateurs maximaux monotones et semi-groupes de contractions dans les espaces de Hilbert*, North-Holland Publishing Co., Amsterdam, 1973. North-Holland Mathematics Studies, No. 5. Notas de Matemática (50).

[11] M. Burger, K. Frick, S. Osher, and O. Scherzer, *Inverse total variation flow*, Multiscale Model. Simul., 6 (2007), pp. 365–395 (electronic).

[12] A. Chambolle and T. Pock, *A first-order primal-dual algorithm for convex problems with applications to imaging*, J. Math. Imaging Vision, 40 (2011), pp. 120–145.

[13] R.H. Chan, Y. Dong, and M. Hintermüller, *An efficient two-phase $L^1$-TV method for restoring blurred images with impulse noise*, IEEE Trans. Image Process., 19 (2010), pp. 1731–1739.

[14] P.G. Ciarlet, *The Finite Element Method for Elliptic Problems*, North Holland, Amsterdam, 1978.

[15] Ph. Clément, *Approximation by finite element functions using local regularization*, Rev. Française Automat. Informat. Recherche Opérationnelle Sér. RAIRO Analyse Numérique, 9 (1975), pp. 77–84.

[16] D.C. Dobson and C.R. Vogel, *Convergence of an iterative method for total variation denoising*, SIAM J. Numer. Anal., 34 (1997), pp. 1779–1791.

[17] V. Duval, J.-F. Aujol, and Y. Gousseau, *The TVL1 model: a geometric point of view*, Multiscale Model. Simul., 8 (2009), pp. 154–189.

[18] I. Ekeland and R. Temam, *Convex analysis and variational problems*, vol. 28 of Classics in Applied Mathematics, Society for Industrial and Applied Mathematics (SIAM), Philadelphia, PA, english ed., 1999. Translated from the French.

[19] C.M. Elliott and S.A. Smitsmeian, *Numerical analysis of the TV regularization and $H^{-1}$ fidelity model for decomposing an image into cartoon plus texture*, IMA J. Numer. Anal., 29 (2009), pp. 651–689.

[20] A. Ern and J.-L. Guermond, *Theory and practice of finite elements*, vol. 159 of Applied Mathematical Sciences, Springer-Verlag, New York, 2004.

[21] X. Feng and A. Prohl, *Analysis of total variation flow and its finite element approximations*, M2AN Math. Model. Numer. Anal., 37 (2003), pp. 533–556.

[22] X. Feng, M. von Oehsen, and A. Prohl, *Rate of convergence of regularization procedures and finite element approximations for the total variation flow*, Numer. Math., 100 (2005), pp. 441–456.

[23] P.-W. Fok, R.R. Rosales, and D. Margetis, *Facet evolution on supported nanostructures: Effect of finite height*, Phys. Rev. B, 78 (2008), p. 235401.

[24] G.B. Folland, *Real analysis*, Pure and Applied Mathematics (New York), John Wiley & Sons Inc., New York, second ed., 1999. Modern techniques and their applications, A Wiley-Interscience Publication.

[25] E.S. Fu, D.-J. Liu, M.D. Johnson, J.D. Weeks, and E.D. Williams, *The effective charge in surface electromigration*, Surface Science, 385 (1997), pp. 259 – 269.

[26] M.-H. Giga and Y. Giga, *Very singular diffusion equations: second and fourth order problems,*
[27] M.-H. Giga, Y. Giga, and R. Kobayashi, Very singular diffusion equations, in Taniguchi Conference on Mathematics Nara ’98, vol. 31 of Adv. Stud. Pure Math., Math. Soc. Japan, Tokyo, 2001, pp. 93–125.

[28] R. Hardt and X. Zhou, An evolution problem for linear growth functionals, Comm. Partial Differential Equations, 19 (1994), pp. 1879–1907.

[29] R. Kobayashi, J.A. Warren, and W.C. Carter, A continuum model of grain boundaries, Physica D: Nonlinear Phenomena, 140 (2000), pp. 141 – 150.

[30] R.V. Kohn and H.M. Versieux, Numerical analysis of a steepest-descent PDE model for surface relaxation below the roughening temperature, SIAM J. Numer. Anal., 48 (2010), pp. 1781–1800.

[31] W.G. Litvinov, T. Rahman, and X.-C. Tai, A modified TV-Stokes model for image processing, SIAM J. Sci. Comput., 33 (2011), pp. 1574–1597.

[32] R.H. Nochetto and A.J. Salgado, A TV diminishing interpolation operator and applications. Submitted, arXiv:1211.1069, 2012.

[33] R.H. Nochetto, G. Savaré, and C. Verdi, A posteriori error estimates for variable time-step discretizations of nonlinear evolution equations, Comm. Pure Appl. Math., 53 (2000), pp. 525–589.

[34] L.I. Rudin, S. Osher, and E. Fatemi, Nonlinear total variation based noise removal algorithms, Physica D: Nonlinear Phenomena, 60 (1992), pp. 259–268.

[35] J. Rulla, Error analysis for implicit approximations to solutions to Cauchy problems, SIAM J. Numer. Anal., 33 (1996), pp. 68–87.

[36] L. Tartar, An introduction to Sobolev spaces and interpolation spaces, vol. 3 of Lecture Notes of the Unione Matematica Italiana, Springer, Berlin, 2007.

[37] J. Wang and B.J. Lucier, Error bounds for finite-difference methods for Rudin-Osher-Fatemi image smoothing, SIAM J. Numer. Anal., 49 (2011), pp. 845–868.

[38] W.P. Ziemer, Weakly differentiable functions, vol. 120 of Graduate Texts in Mathematics, Springer-Verlag, New York, 1989. Sobolev spaces and functions of bounded variation.