Abstract. $L^2$-TGV-regularization has been introduced by Bredies, Kunisch, and Pock [4]. This regularization method requires careful tuning of two regularization parameters. The focus of this paper is to derive analytical results, which allow for characterizing parameter settings, which make this method in fact different from $L^2 - TV$ (the ROF-model) and $L^2 - TV^2$ regularization, respectively. In this paper we also provide explicit solutions of TGV-denoising for particular one-dimensional function data.

1. Introduction

TGV-denoising has been introduced in [4]. Given $k \in \mathbb{N}_0$ and a function $u^\delta : \Omega \to \mathbb{R}$, where $\Omega \subset \mathbb{R}^d$, the method consists in determining

$$u_\lambda := \argmin \left\{ G^k_\lambda(u) : u \in L^2(\Omega) \right\} ,$$

where

$$G^k_\lambda(u) := \frac{1}{2} \int_\Omega (u - u^\delta)^2 \, dx + TGV^k_\lambda(u)$$

with

$$TGV^k_\lambda(u) = \sup \left\{ \int_\Omega u (\nabla \cdot)^k \phi \, dx : \phi \in C^\infty_0(\Omega, \text{Sym}^k(\mathbb{R}^d)), \| (\nabla \cdot)^{k-l} \phi \|_{L^\infty} \leq \lambda_l, \ l = 1, \ldots, k \right\} ,$$

and $\text{Sym}^k(\mathbb{R}^d)$ denotes the space of symmetric tensors of order $k$ with arguments in $\mathbb{R}^d$. There can be imagined several realizations of $\| (\nabla \cdot)^{k-l} \phi \|_{L^\infty}$ to be implemented - one of them is $\| (\nabla \cdot)^{k-l} \phi \|_{L^\infty} = \sup \left\{ \| (\nabla \cdot)^{k-l} \phi(x) \|_{L^2} : x \in \Omega \right\}$, where $\| \cdot \|_{L^2}$ denotes the Frobenius-norm of a tensor. Note that the definition here is slightly different to [4], where in the original definition, the enumeration of the indices of $\lambda_i$ is reversed.

All along this paper, for the simplification of notation and considerations, we restrict attention to the case $k = 2$. Consequently, from now on, we omit the superscript $k$ in the TGV-functional.

The goal of this paper is to increase the knowledge about structural properties of TGV-denoising, and to put this method into perspective with total variation and second order total variation regularization by analytical means. This is done in two different ways:

1. The main result of this paper concerns the characterization of the sets of regularization parameters $\lambda = (\lambda_1, \lambda_2)$ such that the minimizers of TGV$_{\lambda_1,\lambda_2}$ either equal total variation minimizers or minimizers of the second order total variation minimization, and to determine sets of parameters, where TGV minimization is in fact different from either one of them.
2. We study analytical solutions of simple one-dimensional test-cases, where \( d = 1, \Omega = (-1, 1), \) and \( k = 2. \) In this simple situation TGV-denoising (1.1) simplifies to minimizing the functional
\[
G : L^2(-1, 1) \to \mathbb{R} \cup \{+\infty\}, \quad u \to G(u) := \frac{1}{2} \int_{-1}^{1} (u - u^{\delta})^2 \, dx + \text{TGV}_{\lambda_1, \lambda_2}(u). \tag{1.3}
\]
In the specific one-dimensional situation TGV\(_{\lambda_1, \lambda_2}\) can be written as
\[
\text{TGV}_{\lambda_1, \lambda_2}(u) = \sup \left\{ \int_{-1}^{1} u\phi'' \, dx : \phi \in C_c^\infty(-1, 1), \|\phi'\|_{L^\infty} \leq \lambda_1, \|\phi\|_{L^\infty} \leq \lambda_2 \right\}. \tag{1.4}
\]
Similar, as in our previous work [11] for total variation minimization and minimization with totally bounded second derivative, it is possible to characterize the minimizers of TGV\(_{\lambda_1, \lambda_2}\) in a simple manner using Fenchel-duality theory. We show that the minimizers are either equal to \( u^{\delta} \) or piecewise affine linear that bend or jump, whenever the first or second primitives of the dual functions attains an extremum.

We then study explicit solutions of TGV-denoising for the basic test data cases
\[
x \to u^{\delta}(x) = |x| - \frac{1}{2},
\]
\[
x \to u^{\delta}(x) = 1_{[-1/2, 1/2]}(x) - \frac{1}{2},
\]
and
\[
x \to u^{\delta}(x) = x^2 - \frac{1}{3}.
\]
For the first two exemplary cases the minimizers of the TGV-functional (1.3) are weighted sums of TV-minimizers and TV\(^2\)-minimizers. The second example has also been studied in [3] - however, there no complete characterization of the parameter sets have been stated where the TGV\(_{\lambda_1, \lambda_2}\)-minimizer equals either \( L^2 - \text{TV}, L^2 - \text{TV}^2\)-minimizers, which is a focus topic of this work.

The outline of this paper is as follows: In Section 2 we introduce preliminary notation and the main definitions. We derive characteristic properties of minimizers of the TGV-denoising problem (in \( d \)-dimensions) via convex duality theory (Sections 3-4). Later we restrict attention to the case \( d = 1 \) and show that minimizers are either equal to the data or piecewise affine linear (cf. Section 5). Finally we calculate explicit minimizers for the TGV\(_{\lambda_1, \lambda_2}\)-functional in the case where the data are the absolute value (Section 6), the indicator function (Section 7), or a quadratic polynomial (Section 8), respectively.

2. Notation Let \( \Omega \subseteq \mathbb{R}^d \) be a bounded, connected domain with Lipschitzian boundary. Moreover, let \( u^{\delta} : \Omega \to \mathbb{R} \in L^2(\Omega) \). For \( i \in \mathbb{N} \) we define the following functional:
\[
F^i : L^2(\Omega) \to \mathbb{R} \cup \{+\infty\}, \quad u \to F^i(u) := \frac{1}{2} \int_{\Omega} (u - u^{\delta})^2 \, dx + \text{TV}_{\lambda_i}(u). \tag{2.1}
\]
where
\[ TV^i_{\lambda_i}(u) := \sup \left\{ \int_{\Omega} u \phi(i) \, dx : \phi \in C_c^\infty(\Omega, \text{Sym}^k(\mathbb{R}^d)), \left\| (\nabla \cdot)^i \phi \right\|_{L^\infty} \leq \lambda_i \right\}, \tag{2.2} \]

where
\[ \left\| (\nabla \cdot)^k v \right\|_{L^\infty} = \sup \left\{ \left| (\nabla \cdot)^k v(x) \right|_{L^2} : x \in \Omega \right\} , \]
and \( \left| (\nabla \cdot)^i \phi \right|_{L^2} \) denotes the Frobenius-norm of \( (\nabla \cdot)^i \phi \).

The minimizer of (2.1) is denoted by \( v^i_{\lambda_i} \). The minimizer of (1.2) is denoted by \( u_{\lambda_1, \lambda_2} \).

Because
\[ TV^i_{\lambda_i}(u) = \lambda_i TV^i(u), \tag{2.3} \]
we see that minimization of the functional \( F^i \) from (2.1) is standard \( L^2 - TV \)-minimization with regularization parameter \( \lambda_i \). \( L^2 - TV \)-minimization has been studied widely in the literature. In the one-dimensional \( d = 1 \) setting it is used for regression (see e.g. \[8, 6\]) - analytical solutions have been calculated for instance in \[5\]. In image processing, for \( d \geq 2 \), \( L^2 - TV \)-regularization it is called the Rudin-Osher-Fatemi model \[12\]. Regularization with derivatives of higher order bounded variation has been studied for instance in \[14, 10, 11, 13\].

3. Fenchel duality and applications

In the following let \( H \) be a Hilbert-space. In this case it is common to identify \( H \) with its dualspace and to identify the dual pairing \( \langle u^*, u \rangle \) on \( H^* \) and \( H \) with the inner product on \( H \). For instance when \( H = L^2(\Omega), \langle u^*, u \rangle = \int_{\Omega} u^* u \, dx \).

We start by defining the \( * \)-number, which is a generalization of the dualnorm of a Banach-space, to convex, positively homogeneous functionals.

**Definition 3.1.** A proper, convex functional \( T : H \to \mathbb{R} \cup \{+\infty\} \) is positively homogeneous, if there exists some \( l = 1, 2 \ldots \) such that \( T(\lambda u) = |\lambda|^l T(u), \quad \forall \lambda \in \mathbb{R} \).

**Definition 3.2 (The \( * \)-number).** Let \( R : H \to \mathbb{R} \cup \{+\infty\} \) be a positively homogeneous and convex functional. For \( u^* \in H \) define
\[ \| u^* \|_{*, R} := \sup \{ \langle u^*, u \rangle : u \in H, R(u) \leq 1 \} . \]

Moreover define
\[ B^*_R := \left\{ u^* \in H : \| u^* \|_{*, R} \leq 1 \right\} \]
as the dualball with respect to the \( * \)-number.

**Example 3.1.** From (2.3) it follows that
\[ \| u^* \|_{*, TV^i_{\lambda_i}} = \frac{1}{\lambda_i} \| u^* \|_{*, TV^i} \] .
\[ \tag{3.1} \]
Note that according to our definition

\[ TV_i^1 = TV_i^3, \quad \forall i \in \mathbb{N}. \]

**Lemma 3.3.**  
(see [13, Lemma 4.6]). Let \( R \) be positively homogeneous and set

\[ \mathcal{P} := \{ p \in \mathcal{H} : R(p) = 0 \}. \]

From the assumptions that \( R \) is positively homogeneous and convex, it follows that \( \mathcal{P} \) is a linear subspace of \( \mathcal{H} \). Denote by

\[ \mathcal{P}^\perp := \{ u^* \in \mathcal{H} : \langle u^*, p \rangle = 0, p \in \mathcal{P} \}. \]

Then \( \| u^* \|_{\ast, R} = +\infty \) for all \( u^* \notin \mathcal{P}^\perp \).

**Definition 3.4.** Assume that \( i = 1, 2, \ldots \). Let \( \mathcal{H}^i = L^2(\Omega) \), \( R = TV_{\lambda_i} \), and let \( \mathcal{P}^i \) be the set of polynomials of order \( i - 1 \). Then

\[ \mathcal{H}^i := \left\{ u \in L^2(\Omega) : \int_{\Omega} u(x)x^j \, dx = 0, \, |j| = 0, 1, \ldots, i - 1, \, j \in \mathbb{N}_0^d \right\} = \mathcal{P}^{i\perp}. \quad (3.2) \]

**Remark 3.2.** Because \( C^\infty_c(\Omega) \) is dense in \( L^2(\Omega) \), and \( \Omega \) is assumed to be bounded, it follows that

\[ \mathcal{H}^i := \left\{ u \in C^\infty_c(\Omega) : \int_{\Omega} u(x)x^j \, dx = 0, \, |j| = 0, 1, \ldots, i - 1, \, j \in \mathbb{N}_0^d \right\}. \]

The following lemma is a direct consequence of Lemma 3.3 and the above definition:

**Lemma 3.5.**
- \( TV_{\lambda_i}^i, i = 1, 2 \): For all \( u^* \notin \mathcal{H}^i \), \( \| u^* \|_{\ast, TV_{\lambda_i}^i} = +\infty \).
- \( TGV\text{-functional} \): For all \( u^* \notin \mathcal{H}^2 \), \( \| u^* \|_{\ast, TGV_{\lambda_1, \lambda_2}} = +\infty \).

The definitions of the \( \| \cdot \|_{\ast, TV_{\lambda_i}^i} \)-norms are similar as in Meyer’s book [9]. The difference is that there \( \Omega = \mathbb{R}^d \) is considered, and the elements of the space \( L^2(\mathbb{R}^d) \) satisfy natural boundary conditions at \( \infty \). Since we consider bounded domains \( \Omega \) we restrict attention to the subspaces \( \mathcal{H}^i \) rather than to \( L^2 \), as in Meyer’s book. Another possibility, instead of factorizing out polynomials, is to consider boundary conditions on the bounded domain \( \Omega \), which has been realized in [4].

The Fenchel dual of a proper functional \( S : \mathcal{H} \to \mathbb{R} \cup \{ +\infty \} \) is defined as

\[ S^* : \mathcal{H} \to \mathbb{R} \cup \{ +\infty \} \]

\[ u^* \mapsto S^* := \sup_{u \in \mathcal{H}} \{ \langle u^*, u \rangle - S(u) \}. \]

The following results can be found in [7], see also [13]:

**Remark 3.3.**
- Let \( T \) be 1-homogeneous, then the Fenchel dual function is a characteristic function of a convex set \( C^* \). That is,

\[ T^*(u^*) = \chi_{C^*}(u^*) = \begin{cases} 0 & \text{for } u^* \in C^*, \\ +\infty & \text{else}. \end{cases} \quad (3.3) \]
In particular, for $T$ 1-homogeneous,
\[ C^* = B^*_T. \]  

Let $S,R$ be convex and proper functionals defined on $H$. Denote by $\hat{u}$ a minimizer of the functional $u \rightarrow S(u) + R(u)$ and denote by $\hat{u}^*$ a minimizer of the functional $u^* \rightarrow S^*(u^*) + R^*(-u^*)$. Then the extremality conditions holds:
\[ \hat{u}^* \in \partial S(u) \quad \text{and} \quad -\hat{u}^* \in \partial R(u). \]  

**Example 3.4.**

1. The dual functional of
\[ S : L^2(\Omega) \rightarrow \mathbb{R}, \]
\[ u \rightarrow \frac{1}{2} \int_{\Omega} (u - u^\delta)^2 \, dx, \]  
is given by
\[ S^* : L^2(\Omega) \rightarrow \mathbb{R}, \]
\[ u^* \rightarrow \frac{1}{2} \int_{\Omega} u^{*2} \, dx + \int_{\Omega} u^* u^\delta \, dx. \]  
In the case of the quadratic functional the extremality condition (3.5) for a minimizer shows:
\[ \hat{u}^* = \hat{u} - u^\delta. \]  

2. Let
\[ R = TV^\delta_{\lambda_1} : L^2(\Omega) \rightarrow \mathbb{R} \cup \{+\infty\}, \]  
which is 1-homogeneous. Then
\[ R^*(u^*) = \sup_{u \in L^2(\Omega)} \left\{ \int_{\Omega} u^* u \, dx - TV^\delta_{\lambda_1}(u) \right\} = \chi_{B^*_{TV^\delta_{\lambda_1}}}(u^*), \]  
where the characteristic function is 0 on the closed unit ball $C^* = B^*_{TV^\delta_{\lambda}}$ and $+\infty$ else.

3. Let
\[ R = TGV^\delta_{\lambda_1,\lambda_2} : L^2(\Omega) \rightarrow \mathbb{R} \cup \{+\infty\}, \]  
which is 1-homogeneous too. Thus $R^*(u^*) = \chi_{B^*_{TGV^\delta_{\lambda_1,\lambda_2}}}.$

**4. Regularization Methods with 1-Homogeneous Regularizers**

In the following we derive some properties of regularization functionals with $S$ from (3.6) and 1-homogeneous regularizers $R$. We denote by
\[ u \rightarrow E(u) := S(u) + R(u). \]

**Remark 4.1.** From (3.4) we know that
\[ R^* = \chi_{B^*_R}. \]
Then the extremality condition (3.8) guarantees that \( \hat{u}^* = \hat{u} - u^\delta \in B^*_R \) and from Fenchel-duality theory we see that

\[
\begin{align*}
\frac{1}{2} \int_\Omega (\hat{u} - u^\delta)^2 \, dx + R(\hat{u}) \\
= \inf \{ S(u) + R(u) \} \\
= - \inf \{ S^*(u^*) + R^*(-u^*) \} \\
= - S^*(\hat{u}^*) - \chi_{B^*_R}(-\hat{u}^*) \\
= - \int_\Omega \frac{1}{2} (\hat{u} - u^\delta)^2 + u^\delta (\hat{u} - u^\delta) \, dx.
\end{align*}
\]  

(4.1)

In summary we have shown that

\[
\|\hat{u}^*\|_{s, R} \leq 1, \\
R(\hat{u}) = -\int_\Omega (\hat{u} - u^\delta)^2 + u^\delta (\hat{u} - u^\delta) \, dx = -\int_\Omega (\hat{u} - u^\delta) \, d\hat{u}.
\]  

(4.2)

- \boxed{4.2} applied to \( L^2 - TV\)-minimization, shows that \( v^i_{\lambda_i} \), the minimizer of \( F^i \) (see (2.1)), satisfies

\[
\|v^i_{\lambda_i}\|_{s, TV^i_{\lambda_i}} \leq 1 \text{ and } TV^i_{\lambda_i}(v^i_{\lambda_i}) = -\int_\Omega (v^i_{\lambda_i} - u^\delta)v^i_{\lambda_i} \, dx.
\]  

(4.3)

- \boxed{4.2} applied to \( TGV_{\lambda_1, \lambda_2} \)-minimization, shows that \( u_{\lambda_1, \lambda_2} \), the minimizer of \( G \) (1.3), satisfies

\[
\|u^*_{\lambda_1, \lambda_2}\|_{s, TGV^i_{\lambda_1, \lambda_2}} \leq 1 \text{ and } TGV^i_{\lambda_1, \lambda_2}(u_{\lambda_1, \lambda_2}) = -\int_\Omega (u_{\lambda_1, \lambda_2} - u^\delta)u_{\lambda_1, \lambda_2} \, dx.
\]  

(4.4)

Now \( u^*_{\lambda_1, \lambda_2} \in H^2 \) (Lemma 3.3). Since \( u^*_{\lambda_1, \lambda_2} = u_{\lambda_1, \lambda_2} - u^\delta \), \( u^\delta \in H^2 \) implies that also \( u_{\lambda_1, \lambda_2} \in H^2 \), hence by assuming that \( u^\delta \in H^2 \) we can equivalently minimize \( G \) over \( H^2 \) instead of \( L^2(\Omega) \).

**Lemma 4.1.** If \( u \) satisfies \( \boxed{4.2} \) then \( u \) minimizes \( E \).

In particular

- \( u \) minimizes \( F^i \) iff \( -\int_\Omega (u - u^\delta)u \, dx = TV^i_{\lambda_i}(u) \) and \( u - u^\delta \in B^*_{TV^i_{\lambda_i}} \),

- and \( u \) minimizes the \( TGV \)-functional iff \( -\int_\Omega (u - u^\delta)u \, dx = TGV^i_{\lambda_1, \lambda_2}(u) \) and \( u - u^\delta \in B^*_{TGV^i_{\lambda_1, \lambda_2}} \).

**Proof.** We prove the lemma by contradiction: Assume that \( u \) satisfies the assumptions of the lemma but is not a minimizer of \( E \). Then there exists some \( v \neq u \) such that \( v \) minimizes \( E \) and \( E(v) < E(u) \). From \( \boxed{4.4} \) it then follows that

\[
R(v) = -\int_\Omega (v - u^\delta)v \, dx.
\]
Therefore, from the assumption that $u$ satisfies (4.2), we see that
\[
\frac{1}{2} \int_\Omega (v - u^\delta)^2 \, dx - \int_\Omega (v - u) \, dx
\]
\[
= \frac{1}{2} \int_\Omega (v - u^\delta)^2 \, dx + R(v)
\]
\[
< \frac{1}{2} \int_\Omega (u - u^\delta)^2 \, dx + R(u)
\]
\[
= \frac{1}{2} \int_\Omega (u - u^\delta)^2 \, dx - \int_\Omega (u - u^\delta) \, dx
\]
such that
\[
- \frac{1}{2} \int_\Omega v^2 \, dx < - \frac{1}{2} \int_\Omega u^2 \, dx.
\] (4.5)

The dual functional of a convex, 1-homogeneous function $R$, is the characteristic function of $B_R^*$ (cf. Remark 4.1). The Fenchel-duality theorem (see e.g. [7]) states, that $v^* := v - u^\delta$ minimizes the functional $w^* \to S^*(w^*)$ over $B_R^*$, where $S^*$ is as in (3.7), such that we have now
\[
S^*(v - u^\delta) = \int_\Omega \left( \frac{1}{2}(v - u^\delta)^2 + (v - u^\delta)u^\delta \right) \, dx
\]
\[
\leq S^*(u - u^\delta) = \int_\Omega \left( \frac{1}{2}(u - u^\delta)^2 + (u - u^\delta)u^\delta \right) \, dx.
\]

The inequality above simplifies to
\[
\frac{1}{2} \int_\Omega v^2 \, dx \leq \frac{1}{2} \int_\Omega u^2 \, dx,
\]
such that we obtain a contradiction to (4.5). Hence the assumption that $v \neq u$ is a minimizer of $E$ was wrong. □

**Lemma 4.2.** Assume that $R$ is 1-homogeneous functional on $H$. Then $u_{\min} \equiv 0$ minimizes $E$ if and only if $\|u^\delta\|_{*,R} \leq 1$.

**Proof.**
- $0$ minimizes $E \Rightarrow \|u^\delta\|_{*,R} \leq 1$: If $u_{\min} \equiv 0$, then $u^*_\min = -u^\delta$ and the extremality conditions from Remark 4.1 state that $u^*_\min \in B_R^*$. This means that $\|u^*_\min\|_{*,R} \leq 1$ and consequently $\|u^\delta\|_{*,R} \leq 1$.
- $\|u^\delta\|_{*,R} \leq 1 \Rightarrow u_{\min} \equiv 0$: We prove this implication by contradiction. Assume therefore that $\|u^\delta\|_{*,R} \leq 1$ and that $u_{\min} \neq 0$ minimizes $E$. This, in particular, means that $R(u_{\min}) < +\infty$. Then from (4.2) it follows that
\[
- \int_\Omega u_{\min}(u_{\min} - u^\delta) \, dx = R(u_{\min}) \geq \|u^\delta\|_{*,R} R(u_{\min}).
\]

Rearranging the terms and division by $R(u_{\min})$ shows that
\[
- \int_\Omega \frac{u_{\min}^2}{R(u_{\min})} \, dx + \int_\Omega \frac{u_{\min}}{R(u_{\min})} u^\delta \, dx
\]
\[
\geq \|u^\delta\|_{*,R}
\]
\[
= \sup \left\{ \int_\Omega u^\delta \phi \, dx : R(\phi) \leq 1 \right\}.\] (4.6)
Since, by assumption, $0 \neq u_{\min} \in L^2(\Omega)$, we also have
\[
\int_{\Omega} \frac{u_{\min}^2}{R(u_{\min})} \, dx > 0.
\]
This, together with (4.6), shows that
\[
\sup \left\{ \int_{\Omega} u^\delta \phi \, dx : \mathcal{R}(\phi) \leq 1 \right\} \geq \left\| u^\delta \right\|_{\mathcal{R}}
\]
\[
= \sup \left\{ \int_{\Omega} u^\delta \phi \, dx : \mathcal{R}(\phi) \leq 1 \right\},
\]
hence we obtain a contradiction to the assumption $u_{\min} \neq 0$.

\[\square\]

**Example 4.2.**

1. Let $\mathcal{R} = TV_{\lambda_i}^\delta$, then from Lemma 4.2 it follows that $v_{\lambda_i}^i \equiv 0$ if and only if $\left\| u^\delta \right\|_{TV_{\lambda_i}^\delta} \leq 1$.

2. $TGV_{\lambda_1,\lambda_2}$-minimization: Choose $\mathcal{R} = TGV_{\lambda_1,\lambda_2}$, then from Lemma 4.2 it follows that $u_{\lambda_1,\lambda_2} \equiv 0$ if and only if $\left\| u^\delta \right\|_{TGV_{\lambda_1,\lambda_2}} \leq 1$.

These results are similar as in [9], where $TV$-minimization of functions on $\Omega = \mathbb{R}^d$ have been considered.

5. Extremal Properties and Solutions of 1D-TGV

In the following we consider the case $d = 1$ and $\Omega = (-1,1)$. We derive some characteristic properties of the minimizers $u_{\lambda_1,\lambda_2}$ of the $TGV_{\lambda_1,\lambda_2}$-functional $G_{\lambda_1,\lambda_2}$, defined in (1.2).

Below, by some basic considerations, it is possible to identify sets of parameters $\bar{\lambda} = (\lambda_1, \lambda_2)$ for which $u_{\lambda_1,\lambda_2}$ equals some $v_{\lambda_i}^i$, $i = 1, 2$.

For $d = 1$, the dual-norm $\left\| \cdot \right\|_{TV_{\lambda_i}^\delta}, i = 1, 2, \left\| u^* \right\|_{TGV_{\lambda_1,\lambda_2}}$, respectively, can be easily calculated via integration: To see this, let
\[
\sigma^0[u^*](x) := u^*(x),
\]
\[
\sigma^1[u^*](x) := \int_{-1}^{x} u^*(t) \, dt,
\]
\[
\sigma^{i+1}[u^*](x) := \int_{-1}^{x} \sigma^i[u^*](t) \, dt, \quad \forall u^* \in L^2(-1,1).
\]

**Lemma 5.1.** Let $\Omega = (-1,1)$. Then for all $i = 1, 2, \ldots$,
\[
\Psi^i := \left\{ \psi \in C_c^\infty(-1,1) : \sigma^j[\psi](1) = 0, j = 1, 2, \ldots, i-1 \right\} = \mathcal{H}^i \cap C_c^\infty(-1,1).
\]

Moreover,
\[
\overline{\Psi^i} = \mathcal{H}^i.
\]
Proof. Let $u^* \in \mathcal{H}^i \cap C_c^\infty(-1, 1)$, then
\[
\int_{-1}^1 u^* x^j \, dx = 0, \quad \forall j = 0, 1, \ldots i - 1.
\]
For fixed $i$ we prove by an inductive argument that for $u^* \in \mathcal{H}^i \cap C_c^\infty(-1, 1)$ also $u^* \in \Psi^i$.

- Let $j = 1$: Then $u^* \in \mathcal{H}^i$ implies that
  \[
  \sigma^j[u^*](1) = \sigma^1[u^*](1) = \int_{-1}^1 u^* \, dx = \int_{-1}^1 u^* 1 \, dx = 0.
  \]

- Let $2 \leq j \leq i - 1$ and assume that $\sigma^k[u^*](1) = 0$ for $k = 0, 1, \ldots, j - 1$. Then
  \[
  \sigma^j[u^*](1) = \int_{-1}^1 \sigma^{j-1}[u^*] \, dx = - \int_{-1}^1 \sigma^{j-2}[u^*] x \, dx + \sigma^{j-1}[u^*](1).
  \]

The right hand side vanishes because $u^* \in \mathcal{H}^i \cap C_c^\infty(-1, 1)$ and the induction assumption.

The reverse direction can be performed with an analogous induction argument.

(5.3) follows from Remark 3.2 and the fact that $\mathcal{H}^i$ is closed in $L^2(-1, 1)$.

Using this lemma we are able to derive a characterization of the TV-seminorm via $\sigma^i$: For all $u \in L^2(-1, 1)$ we have the identity:

\[
\text{TV}^i_1(u) = \sup \left\{ \int_{-1}^1 u \phi^i \, dx : \phi \in C_c^\infty(-1, 1), \|\phi\|_{L^\infty} \leq 1 \right\}
\]

\[
= \sup \left\{ \int_{-1}^1 u \phi^i \, dx : \phi^i \in C_c^\infty(-1, 1), \phi^j(\pm 1) = 0, j = 0, 1, \ldots, i - 1, \|\phi\|_{L^\infty} \leq 1 \right\}
\]

\[
= \sup \left\{ \int_{-1}^1 u \psi \, dx : \psi \in \Psi^i \cap C_c^\infty(-1, 1), \|\sigma^i[\psi]\|_{L^\infty} \leq 1 \right\}
\]

\[
\overset{\text{(5.2)}}{=} \sup \left\{ \int_{-1}^1 u \phi \, dx : \phi \in \mathcal{H}^i \cap C_c^\infty(-1, 1), \|\sigma^i[\phi]\|_{L^\infty} \leq 1 \right\}.
\]

Using (5.3), $\Psi^i = \mathcal{H}^i$, and the fact that $u \to \|\sigma^i[u]\|_{L^\infty}$ is lower semi-continuous with respect to the $L^2$-norm, it follows that

\[
\text{TV}^i_{\lambda_i}(u) = \lambda_i \text{TV}^i_1(u)
\]

\[
= \sup \left\{ \int_{-1}^1 u \phi \, dx : \phi \in \mathcal{H}^i, \|\sigma^i[\phi]\|_{L^\infty} \leq \lambda_i \right\}.
\]  

(5.4)

From [11, Theorem 5.1] we then get an equivalent characterization of $\text{TV}^i_{\lambda_i}$:

\[
\text{TV}^i_1(u) = \sup \left\{ \int_{-1}^1 u \psi \, dx : \psi \in \mathcal{H}^i, \|\psi\|_{*, \text{TV}^i_1} \leq 1 \right\}.
\]

(5.5)
In an analogous ways we can rewrite the TGV\(_{\lambda_1,\lambda_2}\)-functional

\[
TGV_{\lambda_1,\lambda_2}(u) = \sup \left\{ \int_{-1}^{1} u\phi'' \, dx : \phi \in C_c^\infty(-1,1), \|\phi'\|_{L^\infty} \leq \lambda_1, \|\phi\|_{L^\infty} \leq \lambda_2 \right\} \quad (5.6)
\]

\[
= \sup \left\{ \int_{-1}^{1} u\phi \, dx : \phi \in H^2, \|\sigma^1[\phi]\|_{L^\infty} \leq \lambda_1, \|\sigma^2[\phi]\|_{L^\infty} \leq \lambda_2 \right\}.
\]

**Lemma 5.2.** Let \(d = 1\) and \(\Omega = (-1,1)\).

- For \(u \in H^2\) and \(i = 1, 2\), we have
  \[
  TGV_{\lambda_1,\lambda_2}(u) \leq TV_{\lambda_i}(u) \quad (5.7)
  \]

- For \(u^* \in H^2\) and \(i = 1, 2\), we have
  \[
  \|u^*\|_{*,TV_{\lambda_i}} \leq \|\sigma^i[u^*]\|_{L^\infty}, \|u^*\|_{*,TV_{\lambda_i}} \leq \|u^*\|_{*,TGV_{\lambda_1,\lambda_2}} \quad (5.8)
  \]

As a consequence

\[
B^*_{TGV_{\lambda_1,\lambda_2}} \subset B^*_{TV_{\lambda_1}} \cap B^*_{TV_{\lambda_2}}.
\]

On the other hand, if \(u^*\) satisfies

\[
\|\sigma^i(u^*)\|_{L^\infty} \leq \lambda_i, \quad i = 1, 2,
\]

then \(u^* \in B^*_{TGV_{\lambda_1,\lambda_2}}\). (5.9)

Moreover,

\[
TGV_{\lambda_1,\lambda_2}(u) = \sup \left\{ \int_{\Omega} u\psi \, dx : \psi \in H^2, \|\psi\|_{*,TGV_{\lambda_1,\lambda_2}} \leq 1 \right\} \quad (5.10)
\]

**Proof.** First, we note that for every \(\rho \in H^2\)

\[
\int_{\Omega} \rho p \, dx = 0, \quad \forall p \in P^1. \quad (5.11)
\]

- From (5.6) it follows that
  \[
  TGV_{\lambda_1,\lambda_2}(u) = \sup \left\{ \int_{-1}^{1} u\phi \, dx : \phi \in H^2, \|\sigma^1[\phi]\|_{L^\infty} \leq \lambda_1, \|\sigma^2[\phi]\|_{L^\infty} \leq \lambda_2 \right\}
  \leq \sup \left\{ \int_{-1}^{1} u\phi \, dx : \phi \in H^2, \|\sigma^2[\phi]\|_{L^\infty} \leq \lambda_2 \right\}
  = TV_{\lambda_2}(u), \quad \forall u \in H^2.
  \]

Moreover, from (5.5) and (5.11) it follows for all \(u \in H^2\) that

\[
\begin{align*}
  &\sup \left\{ \int_{-1}^{1} u\phi \, dx : \phi \in H^1, \|\sigma^1[\phi]\|_{L^\infty} \leq \lambda_1 \right\} \\
  &= \sup \left\{ \int_{-1}^{1} u\phi \, dx : \phi \in H^1 \cap P^2_\perp, \|\sigma^1[\phi]\|_{L^\infty} \leq \lambda_1 \right\} \\
  &= \sup \left\{ \int_{-1}^{1} u\phi \, dx : \phi \in H^2, \|\sigma^1[\phi]\|_{L^\infty} \leq \lambda_1 \right\} \\
  &= TV_{\lambda_1}(u), \quad \forall u \in H^2.
\end{align*}
\]
Thus

\[ \mathrm{TGV}_{\lambda_1, \lambda_2}(u) \leq \mathrm{TV}_{\lambda_1}^1(u), \quad \forall u \in \mathcal{H}. \]

- Because \( \mathrm{TV}_{\lambda_i} \) and \( \mathrm{TGV}_{\lambda_1, \lambda_2} \) are lower semi-continuous on \( L^2(-1, 1) \) it follows that

\[
\{ u \in \mathcal{H}^i : \mathrm{TV}_{\lambda_i}(u) \leq 1 \} \\
\supset \{ u \in C_c^\infty(-1, 1) \cap \mathcal{H}^i : \mathrm{TGV}_{\lambda_1, \lambda_2}(u) \leq 1 \} \\
= \{ u \in \mathcal{H}^i : \mathrm{TGV}_{\lambda_1, \lambda_2}(u) \leq 1 \}, \quad \forall i = 1, 2. \tag{5.12}
\]

Therefore, from (5.11) it follows that

\[
\| u^* \|_{*, \mathrm{TV}_{\lambda_i}} = \sup \left\{ \int_{-1}^1 u^* u \, dx : u \in \mathcal{H}^i, \mathrm{TV}_{\lambda_i}(u) \leq 1 \right\} \\
\leq \sup \left\{ \int_{-1}^1 u^* u \, dx : u \in \mathcal{H}^2, \mathrm{TGV}_{\lambda_1, \lambda_2}(u) \leq 1 \right\} , \quad \forall u^* \in \mathcal{H}. \tag{5.13}
\]

This, together with (5.12) implies that

\[
\| u^* \|_{*, \mathrm{TV}_{\lambda_1}} = \sup \left\{ \int_{-1}^1 u^* u \, dx : u \in \mathcal{H}^i, \mathrm{TV}_{\lambda_1}(u) \leq 1 \right\} \\
= \| u^* \|_{*, \mathrm{TGV}_{\lambda_1, \lambda_2}}, \quad \forall u^* \in \mathcal{H}^2.
\]

The definition of the *-number shows that

\[
\| u^* \|_{*, \mathrm{TV}_{\lambda_1}} = \sup \left\{ \int_{-1}^1 u^* u \, dx : u \in \mathcal{H}^i, \mathrm{TV}_{\lambda_1}(u) = 1 \right\} .
\]

For all \( u \in \mathcal{H}^i \) satisfying \( \mathrm{TV}_{\lambda_1}(u) = 1 \) we have

\[
1 = \mathrm{TV}_{\lambda_1}(u) = \sup \left\{ \int_{-1}^1 u \phi^* \, dx : \phi \in \mathcal{H}^i, \| \sigma^i(\phi^*) \|_{L^\infty} \leq 1 \right\} \\
= \sup \left\{ \int_{-1}^1 \frac{u \phi^*}{\| \sigma^i(\phi^*) \|_{L^\infty}} \, dx : \phi^* \in \mathcal{H}^i \right\} .
\]

Choosing \( \phi^* = u^* \) then gives

\[
\int_{-1}^1 uu^* \, dx \leq \| \sigma^i(u^*) \|_{L^\infty}, \quad \forall u \in \mathcal{H} \mathrm{with} \, TV_{\lambda_1}(u) = 1.
\]

This shows that

\[
\| u^* \|_{*, \mathrm{TV}_{\lambda_1}} \leq \| \sigma^i(u^*) \|_{L^\infty}, \quad \forall u^* \in \mathcal{H}.
\]
• To prove (5.9), we use the definition of the ∗-norm:

\[
\|u^*\|_{,TGV_{\lambda_1,\lambda_2}} = \sup \left\{ \int_{-1}^{1} u^* u \, dx : u \in H^2, TGV_{\lambda_1,\lambda_2}(u) \leq 1 \right\},
\]

\[
TGV_{\lambda_1,\lambda_2}(u) = \sup \left\{ \int_\Omega u \phi'' \, dx : \phi \in C^\infty_c(1,1), \|\phi''\|_{L^\infty} \leq \lambda_1, \|\phi\|_{L^\infty} \leq \lambda_2 \right\}.
\]

The function \( \phi = \sigma^2[u^*] \) satisfies

\[- u^* = \phi'' \]

and by assumption

\[\|\phi\|_{L^\infty} = \|\sigma^1(u^*)\|_{L^\infty} \leq \lambda_1, \|\phi\|_{L^\infty} = \|\sigma^2(u^*)\|_{L^\infty} \leq \lambda_2,\]

If \( TGV_{\lambda_1,\lambda_2}(u) \leq 1 \), it then follows that

\[\int_{-1}^{1} u u^* \, dx \leq TGV_{\lambda_1,\lambda_2}(u) \leq 1.\]

Taking the supremum with respect to \( u \) then shows that \( \|u^*\|_{,TGV_{\lambda_1,\lambda_2}} \leq 1 \).

On the other hand according to (5.13) \( u^* \in B^\ast_{TGV_{\lambda_1,\lambda_2}} \) implies that

\[
\max\left\{ \|u^*\|_{*,TV_{\lambda_i}}, \|u^*\|_{*,TV_{\lambda_i}^2} \right\} \leq 1.
\]

Because

\[\|u^*\|_{*,TV_{\lambda_i}} = \frac{1}{\lambda_i} \|\sigma^i[u^*]\|_{L^\infty}, \quad \forall u^* \in B^\ast_{TGV_{\lambda_1,\lambda_2}},\]

it then follows that

\[\|\sigma^i[u^*]\|_{L^\infty} \leq \lambda_i, \quad \forall u^* \in B^\ast_{TGV_{\lambda_1,\lambda_2}}, \forall i = 1, 2.\]

• As a consequence

\[
TGV_{\lambda_1,\lambda_2}(u) = \sup \left\{ \int_\Omega u \phi'' \, dx : \phi \in C^\infty_c(\Omega), \|\phi''\|_{L^\infty} \leq \lambda_1, \|\phi\|_{L^\infty} \leq \lambda_2 \right\}.
\]

\[
= \sup \left\{ \int_\Omega u \psi \, dx : \psi \in H^2, \|\psi\|_{*,TGV_{\lambda_1,\lambda_2}} \leq 1 \right\}.
\]

Lemma 5.3. For

\[\lambda_2 \geq \|\sigma^2[v^1_{\lambda_1}]\|_{L^\infty},\]

we have \( v^1_{\lambda_1} = u_{\lambda_1,\lambda_2} \) and \( TGV_{\lambda_1,\lambda_2}(v^1_{\lambda_1}) = TV^2_{\lambda_1}(v^1_{\lambda_1}) \). On the other hand, if

\[\lambda_1 \geq \|\sigma^1[v^2_{\lambda_2}]\|_{L^\infty},\]

then \( v^2_{\lambda_2} = u_{\lambda_1,\lambda_2} \) and \( TGV_{\lambda_1,\lambda_2}(v^2_{\lambda_2}) = TV^2_{\lambda_2}(v^2_{\lambda_2}) \).

Proof. We only prove the first assertion. The proof of the second assertion is analogous, and therefore omitted.

We summarize two properties of \( v^1_{\lambda_1} \):
• By assumption \( \| \sigma^2 [v_{\lambda_1}^*] \|_{L^\infty} \leq \lambda_2 \).
• Since \( v_{\lambda_1}^* \) minimizes (2.1) it follows from (11) that \( \sigma_1 [v_{\lambda_1}^*] \leq \lambda_1 \).

From (5.9) it then follows that

\[
v_{\lambda_1}^* \in B_{TGV_{\lambda_1, \lambda_2}}.
\]

Because \( v_{\lambda_1}^* \) is the minimizer of the TV-functional \( F^1 \) it follows from Lemma 4.1 and (5.4) that

\[
- \int_{-1}^{1} (v_{\lambda_1}^* - u) v_{\lambda_1}^* \, dx = \text{TV}_{\lambda_1}^1 (v_{\lambda_1}^*)
\]

= sup \( \left\{ \int_{-1}^{1} v_{\lambda_1}^* v^* \, dx : v^* \in \mathcal{H}^1 \cap C_c^\infty (-1,1), \| \sigma^1 (v^*) \|_{L^\infty} \leq \lambda_1 \right\} \).

Moreover, since \( v_{\lambda_1}^* \in B_{TGV_{\lambda_1, \lambda_2}} \), we have

\[
0 = TGV_{\lambda_1, \lambda_2} (v_{\lambda_1}^*) = \sup_{u \in L^2 (-1,1)} \left\{ \int_{-1}^{1} v_{\lambda_1}^* u \, dx - TGV_{\lambda_1, \lambda_2} (u) \right\},
\]

and in particular for the test function \( u = -v_{\lambda_1}^* \),

\[- \int_{-1}^{1} (v_{\lambda_1}^* - u) v_{\lambda_1}^* \, dx = - \int_{-1}^{1} v_{\lambda_1}^* v_{\lambda_1}^* \, dx \leq TGV_{\lambda_1, \lambda_2} (-v_{\lambda_1}^*) = TGV_{\lambda_1, \lambda_2} (v_{\lambda_1}^*) .
\]

This, together with Lemma 5.2 and (5.17) shows that

\[- \int_{-1}^{1} (v_{\lambda_1}^* - u) v_{\lambda_1}^* \, dx \leq TGV_{\lambda_1, \lambda_2} (v_{\lambda_1}^*) \leq \text{TV}_{\lambda_1}^1 (v_{\lambda_1}^*) = - \int_{-1}^{1} (v_{\lambda_1}^* - u) v_{\lambda_1}^* \, dx ,
\]

and therefore in particular

\[- \int_{-1}^{1} (v_{\lambda_1}^* - u) v_{\lambda_1}^* \, dx = TGV_{\lambda_1, \lambda_2} (v_{\lambda_1}^*) .
\]

Applying Lemma 4.1 with (5.16) and (5.19) shows that \( v_{\lambda_1}^* \) also minimizes the TGV_{\lambda_1, \lambda_2}-regularization (1.3).

**DEFINITION 5.4.** We define

\[ \Lambda := \{ (\lambda_1, \lambda_2) : 1 < \| \sigma^1 (v_{\lambda_1}^*) \|_{L^\infty} \text{ and } 1 < \| \sigma^2 (v_{\lambda_1}^*) \|_{L^\infty} \} \).

**COROLLARY 5.5.** Let \( (\lambda_1, \lambda_2) \in [\| \sigma^1 (u^d) \|_{L^\infty} + \infty) \times [\| \sigma^2 (u^d) \|_{L^\infty} , +\infty) \), then \( u_{\lambda_1, \lambda_2} \equiv 0 \).

*Proof.* Because of (3.1) and (5.7) we have

\[ \lambda_1 \geq \| \sigma^1 (u^d) \|_{L^\infty} \geq \| u^d \|_{*, \text{TV}_1} \text{ and } \lambda_2 \geq \| \sigma^2 (u^d) \|_{L^\infty} \geq \| u^d \|_{*, \text{TV}_2} .
\]

Therefore, from Example 4.2 and Lemma 4.2 it follows that \( v_{\lambda_1}^2 = v_{\lambda_1}^1 \equiv 0 \).

Using Lemma 5.2 it follows that \( v_{\lambda_2}^2 = v_{\lambda_1}^1 = u_{\lambda_1, \lambda_2} \), and therefore the assertion.

**LEMMA 5.6.** Let \( u_{\lambda_1, \lambda_2}^* \) be the minimizer of \( u^* \to S^*(u^*) + TGV_{\lambda_1, \lambda_2}^* (-u^*) \).
1. Then, on each connected component of

\[ B := \{ x : |\sigma^1[u^*_{\lambda_1,\lambda_2}](x)| < \lambda_1 \text{ and } |\sigma^2[u^*_{\lambda_1,\lambda_2}](x)| < \lambda_2 \}, \]

\[ u_{\lambda_1,\lambda_2}(x) |_{B} \text{ is a polynomial of maximal degree } 1. \]

2. If there exists an interval \( A \) such that either \(|\sigma^1[u^*_{\lambda_1,\lambda_2}](x)| = \lambda_1 \text{ for all } x \in A\) or \(|\sigma^2[u^*_{\lambda_1,\lambda_2}](x)| = \lambda_2 \text{ for all } x \in A\), then \( u_{\lambda_1,\lambda_2} = u^\delta \) on \( A \).

3. **Jump Condition:** If there exists \( x_0 \in (-1, 1) \) and \( \epsilon > 0 \) such that

\[
\sigma^1[u^*_{\lambda_1,\lambda_2}](x_0) = \lambda_1 \quad \text{and} \quad |\sigma^1[u^*_{\lambda_1,\lambda_2}](x)| < \lambda_1, \quad \forall x \in (x_0 - \epsilon, x_0 + \epsilon) \setminus \{x_0\} \quad (5.20)
\]

then there exist constants \( c_1 \), and \( d_1 \leq d_2 \) such that

\[
u_{\lambda_1,\lambda_2}(x) = \begin{cases} 
  c_1x + d_1, & x \in (x_0 - \epsilon, x_0), \\
  c_1x + d_2, & x \in (x_0, x_0 + \epsilon). 
\end{cases} \quad (5.21)
\]

If instead of the first condition in (5.20), \( \sigma^1[u^*_{\lambda_1,\lambda_2}](x_0) = -\lambda_1 \) holds, then \( u_{\lambda_1,\lambda_2} \) satisfies (5.21), but \( d_2 \leq d_1 \).

4. **Bending Condition:** If there exists \( x_0 \in (-1, 1) \) and some \( \epsilon > 0 \) such that

\[
\sigma^2[u^*_{\lambda_1,\lambda_2}](x_0) = \lambda_2, \quad |\sigma^2[u^*_{\lambda_1,\lambda_2}](x)| < \lambda_2, \quad \forall x \in (x_0 - \epsilon, x_0 + \epsilon) \setminus \{x_0\} \quad (5.22)
\]

Then

\[
u_{\lambda_1,\lambda_2}(x) = \begin{cases} 
  c_1x + d_1, & x \in (x_0 - \epsilon, x_0) \\
  c_2x + d_2, & x \in (x_0, x_0 + \epsilon) 
\end{cases} \quad (5.23)
\]

is continuous at \( x_0 \), and \( c_2 \leq c_1 \), where the later condition we refer to negative bending.

If instead of the first condition in (5.22), \( \sigma^2[u^*_{\lambda_1,\lambda_2}](x_0) = -\lambda_2 \) holds, then the function is positively bending, i.e., \( c_1 \leq c_2 \).

**Proof.** Recall that if \( w^* \not\in \mathcal{H}^2 \), then \( w^* \not\in \mathcal{B}^{\ast}_{\text{TGV}_{\lambda_1,\lambda_2}} \), hence in the following, we restrict our attention to \( w^* \in \mathcal{H}^2 \). The Kuhn-Tucker condition \( -u_{\lambda_1,\lambda_2} \in \partial \mathcal{R}^\ast\left(u^*_{\lambda_1,\lambda_2}\right) \) guarantees that:

\[
\mathcal{R}^\ast(v^*) - \mathcal{R}^\ast(u^*_{\lambda_1,\lambda_2}) + \int_{-1}^{1} u_{\lambda_1,\lambda_2}(v^* - u^*_{\lambda_1,\lambda_2}) \, dx \geq 0, \quad \forall v^* \in \mathcal{H}^2.
\]

In particular, for \( w^* \in \mathcal{B}^{\ast}_{\text{TGV}_{\lambda_1,\lambda_2}} \) we have

\[
\int_{-1}^{1} u_{\lambda_1,\lambda_2} \left(w^* - u^*_{\lambda_1,\lambda_2}\right) \, dx \geq 0. \quad (5.24)
\]

**Item (1):** Let \((a, b)\) be an open interval such that

\[
|\sigma^i[u^*_{\lambda_1,\lambda_2}](x)| < \lambda_i, \text{ for all } i = 1, 2 \text{ and } x \in (a, b).
\]
Moreover, let $\phi \in C^\infty_c(-1,1)$ with $\text{supp}(\phi) \subseteq (a,b)$ such that also
\[ |\sigma_i[u_{\lambda_1,\lambda_2}^*(x)] + \phi^{(2-i)}(x)| < \lambda_i, \quad \text{for all } i = 1, 2 \text{ and } x \in (a,b). \]

Then,
\[ w^* := u_{\lambda_1,\lambda_2}^* + \phi'' \in B_{TGV_{\lambda_1,\lambda_2}} \]
and therefore, it follows from (5.24) that
\[ -\int_a^b \phi'' u_{\lambda_1,\lambda_2} \, dx \leq 0, \quad \forall \phi \in C^\infty_c(-1,1) \text{ with } \text{supp}(\phi) \subseteq (a,b). \]

Hence, $u_{\lambda_1,\lambda_2}$ is a polynomial of order one in the interval $(a,b)$.

**Item (2):**

- i) Assume that

  \[ \sigma^1[u_{\lambda_1,\lambda_2}^*(x)] = \lambda_1, \quad \forall x \in (a,b). \]

  Then,
  \[ u^\delta(x) - u_{\lambda_1,\lambda_2}(x) = u_{\lambda_1,\lambda_2}^*(x) = (\sigma^1[u_{\lambda_1,\lambda_2}^*])'(x) = 0, \quad \forall x \in (a,b), \]
  and therefore $u^\delta(x) = u_{\lambda_1,\lambda_2}(x)$ in $(a,b)$.

- ii) Assume that

  \[ |\sigma^2[u_{\lambda_1,\lambda_2}^*(x)]| = \lambda_2, \quad \forall x \in (a,b). \]

  From this it follows that
  \[ 0 = (\sigma^2[u_{\lambda_1,\lambda_2}^*])''(x) 
  = (\sigma^1[u_{\lambda_1,\lambda_2}^*])'(x) 
  = u_{\lambda_1,\lambda_2}(x) - u^\delta(x), \quad \forall x \in (a,b), \]
  and therefore $u^\delta(x) = u_{\lambda_1,\lambda_2}(x)$ in $(a,b)$. 

**Item (4):** Item 4 is based on the Assumption that there exists $\epsilon > 0$ and $x_0 \in (-1, 1)$ such that

$$\sigma^2[u_{\lambda_1, \lambda_2}^*(x_0)] = \lambda_2 \quad \text{and} \quad \sigma^2[u_{\lambda_1, \lambda_2}^*(x_0 + y)] < \lambda_2, \quad \forall y \in (0, \epsilon).$$

Then, from Item 1 it follows that $u_{\lambda_1, \lambda_2}$ is piecewise affine linear in $(x_0 - \epsilon, x_0 + \epsilon)$. To be precise, there exists coefficients $c_1, d_1, c_2, d_2$ such that

$$u_{\lambda_1, \lambda_2}(x) = \begin{cases} c_1 x + d_1 & \forall x \in (x_0 - \epsilon, x_0), \\ c_2 x + d_2 & \forall x \in (x_0, x_0 + \epsilon). \end{cases}$$  \hspace{1cm} (5.25)

We prove the assertion of Item 4 in two steps.

1. Firstly we show that the coefficients of the piecewise polynomial satisfy $c_1 \geq c_2$.
2. Secondly we show that $u_{\lambda_1, \lambda_2}$ is continuous at $x_0$, such that we can conclude that it is bending at $x_0$.

a) To prove the first item, $c_1 \geq c_2$, we use some $w^* \in B^*_{TGV_{\lambda_1, \lambda_2}}$ (see Figure 5.2) satisfying

$$\sigma^1[w^*](x_0) = 0,$$  \hspace{1cm} (5.26)

$$\sigma^1[w^*](x) = \sigma^1[u_{\lambda_1, \lambda_2}^*(x)], \quad \forall x \notin (x_0 - \epsilon, x_0 + \epsilon),$$

$$\sigma^1[w^*](x) \leq \sigma^1[u_{\lambda_1, \lambda_2}^*(x)], \quad \forall x \in (x_0 - \epsilon, x_0),$$

$$\sigma^1[w^*](x) \geq \sigma^1[u_{\lambda_1, \lambda_2}^*(x)], \quad \forall x \in (x_0, x_0 + \epsilon),$$

$$\mu := -\int_{x_0 - \epsilon}^{x_0} (\sigma^1[w^*] - \sigma^1[u_{\lambda_1, \lambda_2}^*]) \, dx,$$  \hspace{1cm} (5.27)

and

$$\sigma^2[w^*](x) < \sigma^2[u_{\lambda_1, \lambda_2}^*(x)], \quad \forall x \in (x_0 - \epsilon, x_0 + \epsilon) \setminus \{x_0\},$$

$$0 < \sigma^2[u_{\lambda_1, \lambda_2}^*(x_0) - \sigma^2[w^*](x_0)] < \lambda_2.$$  \hspace{1cm} (5.28)

With such a function $w^*$ it follows from (5.24) that

$$0 \leq \int_{-1}^{1} u_{\lambda_1, \lambda_2} (w^* - u_{\lambda_1, \lambda_2}^*) \, dx$$

$$= -\int_{x_0 - \epsilon}^{x_0 + \epsilon} u_{\lambda_1, \lambda_2}^{\epsilon} (\sigma^1[w^*] - \sigma^1[u_{\lambda_1, \lambda_2}^*]) \, dx$$

$$= c_1 \int_{x_0 - \epsilon}^{x_0} (\sigma^1[w^*] - \sigma^1[u_{\lambda_1, \lambda_2}^*]) \, dx$$

$$\quad - \mu$$

$$= c_2 \int_{x_0}^{x_0 + \epsilon} (\sigma^1[w^*] - \sigma^1[u_{\lambda_1, \lambda_2}^*]) \, dx$$

$$\quad + \mu$$

$$= \mu(c_2 - c_1),$$

which shows that $c_1 \leq c_2$ since $\mu \geq 0$.  \hspace{1cm} (5.29)
b) To prove the continuity of $u_{\lambda_1, \lambda_2}$ we use a function $w^* \in B_{TGV_{\lambda_1, \lambda_2}}$ which satisfies:

\[
s^2[w^*](x) = s^2[u^*_{\lambda_1, \lambda_2}](x) , \quad \forall x \notin (x_0 - \epsilon, x_0 + \epsilon) \tag{5.29a}
\]

\[
s^2[w^*](x) \neq s^2[u^*_{\lambda_1, \lambda_2}](x) , \quad \text{for a. a.} \ x \in (x_0 - \epsilon, x_0 + \epsilon) \tag{5.29b}
\]

\[
s^1[w^* - u^*](x_0) = a \neq 0 , \tag{5.29c}
\]

\[
\int_{x_0 - \epsilon}^{x_0} (s^1[w^*] - s^1[u^*_{\lambda_1, \lambda_2}]) \, dx = - \int_{x_0}^{x_0 + \epsilon} (s^1[w^*] - s^1[u^*_{\lambda_1, \lambda_2}]) \, dx . \tag{5.29d}
\]

Such a function is represented in Figure 5.3. With such a function $w^*$ it follows from (5.24), (5.25), and integration by parts, that

\[
0 \leq \int_{-1}^{1} u_{\lambda_1, \lambda_2} (w^* - u^*_{\lambda_1, \lambda_2}) \, dx
\]

\[
= s^1[w^* - u^*_{\lambda_1, \lambda_2}](x_0) (-c_1 x_0 - d_1 + c_2 x_0 + d_2) = a
\]

\[
- c_1 \int_{x_0 - \epsilon}^{x_0} (s^1[w^*] - s^1[u^*_{\lambda_1, \lambda_2}]) \, dx
\]

\[
- c_2 \int_{x_0}^{x_0 + \epsilon} (s^1[w^*] - s^1[u^*_{\lambda_1, \lambda_2}]) \, dx . \tag{5.30}
\]
Choosing \( \tilde{w}^* \) such that \( \sigma^1[\tilde{w}^* - u^*_{\lambda_1, \lambda_2}](x_0) = -a \), but otherwise satisfying the same properties as \( w^* \), that are (5.29a) and (5.29b), then we obtain

\[
0 \leq \int_{-1}^{1} u_{\lambda_1, \lambda_2} (\tilde{w}^* - u^*_{\lambda_1, \lambda_2}) \, dx
= \sigma^1[\tilde{w}^* - u^*_{\lambda_1, \lambda_2}](x_0) (-c_1 x_0 - d_1 + c_2 x_0 + d_2) - c_1 \int_{x_0-\epsilon}^{x_0} (\sigma^1[\tilde{w}^*] - \sigma^1[u^*_{\lambda_1, \lambda_2}] ) \, dx
- c_2 \int_{x_0}^{x_0+\epsilon} (\sigma^1[\tilde{w}^*] - \sigma^1[u^*_{\lambda_1, \lambda_2}] ) \, dx .
\]

(5.31)

Combining (5.31) and (5.30) finally shows

\[
a (-c_1 x_0 - d_1 + c_2 x_0 + d_2) \leq 0
\]

such that we conclude that \(-c_1 x_0 - d_1 + c_2 x_0 + d_2 = 0\), which shows that \( u_{\lambda_1, \lambda_2} \) is continuous at \( x_0 \).

**Item (3):** Assume that \( u_{\lambda_1, \lambda_2} \) is as in (5.23). In the case where

\[
\sigma^1[u^*_{\lambda_1, \lambda_2}](x_0) = \lambda_1 \quad \text{and} \quad \sigma^1[u^*_{\lambda_1, \lambda_2}](y) < \lambda_1, \quad \forall y \in (x_0 - \epsilon, x_0 + \epsilon), \setminus \{x_0\}
\]

we select some \( w^* \in \mathcal{H}^2 \) such that

\[
u^*_{\lambda_1, \lambda_2}(x) = w^*(x), \quad \forall x \not\in (x_0 - \epsilon, x_0 + \epsilon),
\]

\[
u^*_{\lambda_1, \lambda_2}(x) < w^*(x), \quad \forall x \in (x_0 - \epsilon, x_0), \quad \text{(5.32)}
\]

\[
u^*_{\lambda_1, \lambda_2}(x) > w^*(x), \quad \forall x \in (x_0, x_0 + \epsilon), \quad \text{(5.33)}
\]

and

\[
\sigma^1[u^*_{\lambda_1, \lambda_2}](x_0) = \sigma^1[w^*](x_0) . \quad \text{(5.34)}
\]

Defining

\[
a := \int_{x_0-\epsilon}^{x_0} \sigma^1[w^* - u^*_{\lambda_1, \lambda_2}] \, dx - \int_{x_0}^{x_0+\epsilon} \sigma^1[w^* - u^*_{\lambda_1, \lambda_2}] \, dx ,
\]

(5.24) can be rewritten as

\[
0 \geq \int_{-1}^{1} u_{\lambda_1, \lambda_2} (w^* - u^*_{\lambda_1, \lambda_2}) \, dx
= -(c_1 x_0 + d_1) \sigma^1[w^* - u^*](x_0) - c_1 a
+ (c_2 x_0 + d_2) \sigma^1[w^* - u^*](x_0) + c_2 a
= a(c_2 - c_1) .
\]
Now replacing conditions (5.32), (5.33) by
\[ u^*_1,\lambda_2(x) > w^*(x), \quad \forall x \in (x_0 - \epsilon, x_0), \]
\[ u^*_1,\lambda_2(x) < w^*(x), \quad \forall x \in (x_0, x_0 + \epsilon), \]
and again using (5.24) we also obtain \( a(c_1 - c_2) \leq 0. \) Thus \( c_1 = c_2. \)
(jumping down when \( \sigma_d \) items, we can also proof that
\[ \frac{\partial}{\partial x} \frac{\partial^2}{\partial x^2} u^*_1,\lambda_2(x) = +\lambda_1. \]
Using the same arguments as in previous
items, we can also proof that \( d_1 \geq d_2. \)

LEMMA 5.7. Let \( u_{\lambda_1,\lambda_2} \) be the minimizer of \( u \to S(u) + \text{TGV}_{\lambda_1,\lambda_2}(u). \)

1. If there exists \( x_0 \in (-1, 1), \) such that \( u_{\lambda_1,\lambda_2} \) is as in (5.21) (jumping up
\( d_1 \leq d_2), \) then \( \sigma^1 [u^*_1,\lambda_2] (x_0) = \lambda_1. \)

2. If there exists \( x_0 \in (-1, 1), \) such that \( u_{\lambda_1,\lambda_2} \) is as in (5.23) and \( c_2 \leq c_1 \)
(negative bending), then \( \sigma^2 [u^*_1,\lambda_2] (x_0) = \lambda_2. \)

3. If there exists an interval \( A \) such that \( u_{\lambda_1,\lambda_2}(x) = w^\delta(x) \) for \( x \in A, \) then one
of the two statements holds
\[ (a) \quad \sigma^1 [u^*_1,\lambda_2] (x) = \lambda_1 \text{ for } x \in A, \text{ or} \]
\[ (b) \quad \sigma^2 [u^*_1,\lambda_2] (x) = \lambda_2 \text{ and } \sigma^1 [u^*_1,\lambda_2] (x) = 0 \text{ for } x \in A. \]

Proof. Recall that if \( u_{\lambda_1,\lambda_2} \) is different from \( w^\delta, \) then \( u_{\lambda_1,\lambda_2} \) is a polynomial
(piecewise). Set \( I := (x_0 - \epsilon, x_0 + \epsilon). \)

1. Now assume that \( u_{\lambda_1,\lambda_2} \) as in (5.21) and \( \sigma^1 [u^*_1,\lambda_2] (x) < \lambda_1 \text{ for } x \in I. \)
Then we can find \( w^* \in B^*_1 \lambda_1 \cap B^*_2 \lambda_2 \) such that
\[ w^*(x) = u^*_1,\lambda_2(x) \quad \text{ for } x \in (-1, 1) \setminus I, \]
\[ \sigma^1 [w^*] (x) > \sigma^1 [u^*_1,\lambda_2] (x_0), \]
\[ \sigma^2 [w^*] (x) = \sigma^2 [u^*_1,\lambda_2] (x) \quad \text{ for } x \in (-1, 1) \setminus I. \]
The last condition implies that
\[ \int_{x_0-\epsilon}^{x_0+\epsilon} x (u^*_1,\lambda_2 - w^*) \, dx = 0 \]
such that
\[ \int_{-1}^{1} u_{\lambda_1,\lambda_2}(u^*_1,\lambda_2 - w^*) \, dx = d_1 \int_{x_0-\epsilon}^{x_0+\epsilon} (u^*_1,\lambda_2 - w^*) \, dx + d_2 \int_{x_0}^{x_0+\epsilon} (u^*_1,\lambda_2 - w^*) \, dx \]
\[ = (d_2 - d_1) \left( \sigma^1 [w^*] (x_0) - \sigma^1 [u^*_1,\lambda_2] (x_0) \right) > 0. \]

Now this, together with (4.4), would give
\[ \text{TGV}_{\lambda_1,\lambda_2}(u_{\lambda_1,\lambda_2}) = - \int_{-1}^{1} u_{\lambda_1,\lambda_2}(u^*_1,\lambda_2) \, dx < - \int_{-1}^{1} u_{\lambda_1,\lambda_2} w^* \, dx, \]
which contradicts the definition of TGV_{\lambda_1,\lambda_2} as the supremum of such integrals. Hence \( \sigma^1 [u^*_1,\lambda_2] \) must be maximal at \( x_0. \)
2. Set $I := (x_0 - \epsilon, x_0 + \epsilon)$ and assume that $u_{\lambda_1, \lambda_2}$ is as in (5.23) and

$$\sigma^2 \left[ u_{\lambda_1, \lambda_2}^* \right] (x) < \lambda_2 \text{ for } x \in I.$$ Then we can find $w^* \in B^*_{{\text{TGV}}_{\lambda_1, \lambda_2}}$ such that

$$w^*(x) = u_{\lambda_1, \lambda_2}^*(x) \quad \text{ for } x \in (-1, 1) \setminus I$$

$$\sigma^2 [w^*](x_0) > \sigma^2 \left[ u_{\lambda_1, \lambda_2}^* \right] (x_0)$$

$$\sigma_1 [w^*] (x) = \sigma_1 \left[ u_{\lambda_1, \lambda_2}^* \right] (x) \quad \text{ for } x \in (-1, 1) \setminus I.$$

The last condition and the continuity of $u_{\lambda_1, \lambda_2}$ at $x_0$ imply that

$$\int_{x_0 - \epsilon}^{x_0 + \epsilon} u_{\lambda_1, \lambda_2}^* (w^* - w^*) \, dx = -\int_{x_0 - \epsilon}^{x_0 + \epsilon} u_{\lambda_1, \lambda_2}^* \sigma_1 \left[ u_{\lambda_1, \lambda_2}^* - w^* \right] \, dx$$

$$= -c_1 \int_{x_0 - \epsilon}^{x_0 + \epsilon} \sigma_1 \left[ u_{\lambda_1, \lambda_2}^* - w^* \right] \, dx - c_2 \int_{-\epsilon}^{x_0 + \epsilon} \sigma_1 \left[ u_{\lambda_1, \lambda_2}^* - w^* \right] \, dx$$

such that

$$\int_{-1}^{1} u_{\lambda_1, \lambda_2}^* (w^* - w^*) \, dx = (c_2 - c_1) \left( \sigma^2 \left[ u_{\lambda_1, \lambda_2}^* - w^* \right] (x_0) \right) > 0.$$

Now this, together with (4.4), would give

$$\text{TGV}_{\lambda_1, \lambda_2} (u_{\lambda_1, \lambda_2}) = -\int_{-1}^{1} u_{\lambda_1, \lambda_2}^* w^* \, dx < -\int_{-1}^{1} u_{\lambda_1, \lambda_2}^* w^* \, dx,$$

which contradicts the definition of $\text{TGV}_{\lambda_1, \lambda_2}$ as a supremum. Hence

$$\sigma_1 \left[ u_{\lambda_1, \lambda_2}^* \right]$$

must be maximal at $x_0$.

3. The proof is analog to (1),(2).

\[\square\]

6. Example 1 In the following we calculate the specific minimizers of TV, TV$^2$ and $\text{TGV}_{\lambda_1, \lambda_2}$-minimization for the test data,

$$u^\delta : (-1, 1) \to \mathbb{R}$$

$$x \mapsto |x| - \frac{1}{2}. \quad \text{(6.1)}$$

In this case we have

$$\lambda_1 \|u^\delta\|_{*, \text{TV}^1_{\lambda_1}} = \|u^\delta\|_{*, \text{TV}^1_1} = \|\sigma_1 [u^\delta]\|_L^\infty = \int_{-1}^{-\frac{1}{2}} u^\delta \, dx = \frac{1}{8}$$

and

$$\lambda_2 \|u^\delta\|_{*, \text{TV}^2_{\lambda_2}} = \|u^\delta\|_{*, \text{TV}^2_1} = \|\sigma^2 [u^\delta]\|_L^\infty = |\sigma^2 [u^\delta](0)| = \frac{1}{12} = \frac{2}{3} \|u^\delta\|_{*, \text{TV}^1_1}.$$
6.1. TV-minimizer

Using the same methods as in [11], we find that for given data (6.1), the minimizer of the TV-functional $F^1$ is given by

$$v^1_{\lambda_1} = \begin{cases} \frac{\sqrt{2\lambda_1} - 1}{2} & |x| \leq \sqrt{2\lambda_1} \\ u^\delta(x) & \sqrt{2\lambda_1} < |x| \leq 1 - \sqrt{2\lambda_1} \\ \frac{1}{2} - \frac{1}{2\lambda_1} & 1 - \sqrt{2\lambda_1} < |x| \leq 1 \end{cases}$$

(see Figure 6.2 (right)).

The function $v^1_{\lambda_1}$ and its dual $v^1_{\lambda_1}^*$ satisfy the following properties:

1. $TV(v^1_{\lambda_1}) = 2 - 4\sqrt{2\lambda_1}$,
2. $\|v^1_{\lambda_1}^*\|_{*,TV^1_1} = \sigma^1\left[\|v^1_{\lambda_1}\|_{*,TV^1_1}\right]\left(\frac{1}{2}\right) = \lambda_1$, and
3. $\|v^1_{\lambda_1}^*\|_{*,TV^2_1} = \|\sigma^2\left[v^1_{\lambda_1}^*\right](0)\| = \lambda_1\left(1 - \frac{2}{3}\sqrt{2\lambda_1}\right)$.

Hence, from Lemma 6.3 it follows that, as long as

$$\lambda_2 \geq \|v^1_{\lambda_1}^*\|_{*,TV^2_1} = \lambda_1\left(1 - \frac{2}{3}\sqrt{2\lambda_1}\right),$$

(6.2)

the TV$^1$-minimizer is also the TGV$^{\lambda_1,\lambda_2}$-minimizer and

$$\lambda_1 TV^1_1(v^1_{\lambda_1}) = TV^1_1(v^1_{\lambda_1}) = TGV^1_1(v^1_{\lambda_1}) = \lambda_1(2 - 4\sqrt{2\lambda_1}).$$

6.2. TV$^2$-minimizers

For $u^\delta$ from (6.1) the minimizer of $F^2$ is given by

$$v^2_{\lambda_2} = \left(1 - \frac{1}{\|u^\delta\|_{*,TV^2_1}}\lambda_2\right)^+ u^\delta,$$
where
\[ f^+(x) = \max \{ f(x), 0 \} \]

Using Lemma 5.3 it follows that for
\[ \lambda_1 \geq \| v^2_{\lambda_1} \|_{TV^i} = \left( 1 - \frac{1}{\| u^d \|_{TV^i}} \lambda_2 \right) \| u^d \|_{TV^i}, \] (6.3)

the $TV^2$-minimizer, i.e., the minimizer of $F^2$, is also a minimizer of the $TGV_{\lambda_1,\lambda_2}$-functional $G$. In Figure 6.1 we illustrate the $(\lambda_1, \lambda_2)$-region where the minimizers of $G$ are equal to minimizers of $F^2$.

### 6.3. TGV$_{\lambda_1,\lambda_2}$-minimizer

Firstly, we calculate the set $\Lambda$ (cf. Definition 5.4) for which the $TGV_{\lambda_1,\lambda_2}$-minimizer is different from the $TV_i$-minimizers, respectively. For this particular data $u^\delta$ this means that for
\[ \lambda_2 \notin \Lambda_2 := \left[ \frac{1}{12} - \frac{2}{3} \lambda_1, \left( 1 - \frac{2}{3} \sqrt{2\lambda_1} \right) \lambda_1 \right], \]

the minimizer of the $TGV_{\lambda_1,\lambda_2}$-functional $G$ equals a minimizer of a $TV^i$-functional $F^i$, for some $i = 1, 2$.

Let now $\lambda_2 \in \Lambda_2$, which is the only case for which we can expect that the $TGV_{\lambda_1,\lambda_2}$-minimizer is different to $TV_i$-minimizers.

We introduce the two-parametric set of functions $W$, consisting of all functions of the form,
\[ w(x, c, d) := \begin{cases} 
  d|x| + c(1-d) - \frac{1}{2} & |x| \leq c \\
  u^d & c < |x| \leq 1 - c \\
  d|x| + c(d-1) - d + \frac{1}{2} & |x| > 1 - c
\end{cases}, \] (6.4)

where $c \in [0, \frac{1}{2}]$ and $d \in [0, 1]$. Note that
- $w(x, c, d)$ is continuous,
- $w(x, 0, 1) = w(x, \frac{1}{2}, 1) = u^d(x)$ and $w(x, \frac{1}{2}, 0) = 0$,
- $w(x, c, 0) = v_{\lambda_1}^1(x)$ for $\lambda_1 = \frac{1}{2} c^2$,
- $w(x, \frac{1}{2}, d) = v_{\lambda_2}^2(x)$ for $\lambda_2 = (1-d) \| u^d \|_{TV^i}$.

Assuming that $w_{\lambda_1,\lambda_2} := w(\cdot, c_{\lambda_1,\lambda_2}, d_{\lambda_1,\lambda_2})$ minimizes $G$, Lemma 5.7 provides necessary criteria for optimality of the parameters $c_{\lambda_1,\lambda_2}$ and $d_{\lambda_1,\lambda_2}$, which are derived in the following. Then, in Theorem 6.1 below, we prove that $w_{\lambda_1,\lambda_2}$ in fact minimizes $G$.

Assuming that $w_{\lambda_1,\lambda_2}$ is a minimizer of $G$ it follows from Lemma 5.7 that:
• For \((\lambda_1, \lambda_2)\), such that \(d_{\lambda_1, \lambda_2} > 0\), \(w_{\lambda_1, \lambda_2}\) bends at \(x = 0\). In Remark 6.1, we calculate the coefficients such that \(d_{\lambda_1, \lambda_2} = 0\). Lemma 5.7 item 2 states that then \(w_{\lambda_1, \lambda_2} = w_{\lambda_1, \lambda_2} - u^6\) satisfies

\[
\sigma^2 [w_{\lambda_1, \lambda_2}^*](0) = -\lambda_2. \tag{6.5}
\]

• Lemma 5.7 item 3a states that since \(|\lambda|\leq\lambda\), for given \(\lambda\)

\[\lambda_1 = |\sigma^1 [w_{\lambda_1, \lambda_2}^*](\lambda_1, \lambda_2)| = |\sigma^1 [w_{\lambda_1, \lambda_2}^*](1 - \lambda_1, \lambda_2)|. \tag{6.6}
\]

(Item 3b) cannot occur in this case, because \(\sigma^1[w_{\lambda_1, \lambda_2}^*](-1 + c_{\lambda_1, \lambda_2} \neq 0, \text{ for any } d_{\lambda_1, \lambda_2} \neq 1)\).

Using a Computer Algebra system, we solve (6.5)-(6.6) and obtain

\[
c_{\lambda_1, \lambda_2} = \frac{3 (\lambda_1 - \lambda_2)}{2 \lambda_1}, \quad d_{\lambda_1, \lambda_2} = 1 - \frac{8}{9} \frac{\lambda_2^2}{(\lambda_2 - \lambda_1)^2} \lambda_1. \tag{6.7}
\]

**Remark 6.1.** We want to see what happens for the special case when \((\lambda_1, \lambda_2) \in \partial \Lambda\), that is we consider the two sets of parameters:

\[
\begin{align*}
&\{(\lambda_1, \|v_{\lambda_1}^1\|_{TV^1}), \lambda_1 \in (0, \|u^6\|_{TV^1})\}, \\
&\{(\|v_{\lambda_2}^2\|_{TV^1}, \lambda_2), \lambda_2 \in (0, \|u^6\|_{TV^1})\}.
\end{align*}
\]

• In the case \(\lambda_2 = \lambda_1 (1 + \frac{2}{3} \sqrt{2\lambda_1})\) (this is the case where \(\|v_{\lambda_1}^1\|_{TV^2} = 1\), see Lemma 5.3). (6.7) gives \(d_{\lambda_1, \lambda_2} = 0\) and \(c_{\lambda_1, \lambda_2} = \sqrt{2\lambda_1}\). One can see that then \(w_{\lambda_1, \lambda_2}\) is either piecewise constant or equal to \(u^6\) on \((-1 + c_{\lambda_1, \lambda_2}, -c_{\lambda_1, \lambda_2}) \cup (c_{\lambda_1, \lambda_2}, 1 - c_{\lambda_1, \lambda_2})\). We see that for this particular choice of \((\lambda_1, \lambda_2)\) we have \(w_{\lambda_1, \lambda_2} = v_{\lambda_1}^1\), hence \(w_{\lambda_1, \lambda_2}\) also minimizes \(\mathcal{F}_1\).

• For \(\lambda_1 = \frac{4}{3} \lambda_2\) (this is the case where \(\|v_{\lambda_2}^2\|_{TV^1} = \lambda_1\), see Lemma 5.3), we have \(c_{\lambda_1, \lambda_2} = \frac{1}{2}\) and \(d_{\lambda_1, \lambda_2} = \left(1 - \frac{3}{3} \frac{\lambda_2}{\|u^6\|_{TV^1}}\right)\). We see that \(w_{\lambda_1, \lambda_2} = v_{\lambda_2}^2\).

**Theorem 6.1.** For \((\lambda_1, \lambda_2) \in \Lambda\) and \(c_{\lambda_1, \lambda_2}, d_{\lambda_1, \lambda_2}\) satisfying (6.7), \(w_{\lambda_1, \lambda_2} = w_{\lambda_1, \lambda_2}\).

In order to prove the theorem in a compact way, we need the following remark:

**Remark 6.2.** In the next two items, we only rewrite \(w_{\lambda_1, \lambda_2}\) as a linear combination of minimizers of \(\mathcal{F}_i\), where we have to replace \(\lambda_i\) by a different parameter \(\mu_i\) depending on \(\lambda_1, \lambda_2\).

• For given \(\mu_1 \in [0, \|u^6\|_{TV^1}]\), \(\mu_2 \in [0, \|u^6\|_{TV^1}]\) set

\[
\begin{align*}
\lambda_1 &= \frac{\mu_1 \mu_2}{\|u^6\|_{TV^1}}, \\
\lambda_2 &= \frac{\mu_2}{\|u^6\|_{TV^1}} \|v_{\mu_1}^1\|_{TV^1} = 12 \mu_2 \left(\mu_1 - \frac{2}{3} \sqrt{2\mu_1^3}\right). \tag{6.8}
\end{align*}
\]

Comparing the coefficients of the piecewise terms of \(w_{\lambda_1, \lambda_2}\), we see that for \((\lambda_1, \lambda_2) \in \Lambda\) we can write

\[
w_{\lambda_1, \lambda_2} = \frac{\mu_2}{\|u^6\|_{TV^1}} v_{\mu_1}^1 + v_{\mu_2}^2 \tag{6.9}
\]
• On the other hand, for \((\lambda_1, \lambda_2) \in \Lambda\) given, we calculate \(\mu_1, \mu_2\) by
\[
\mu_2 = \frac{2}{27} \frac{\lambda^2_1}{(\lambda_1 - \lambda_2)^2}, \quad \mu_1 = \frac{9}{8} \frac{(\lambda_1 - \lambda_2)^2}{\lambda^2_1}
\]
and express \(w_{\lambda_1, \lambda_2}\) by (6.9).

Proof. Using the triangle-inequality and the estimate \(\text{TGV}_{\lambda_1, \lambda_2}(u) \leq \lambda_1 TV^i(u)\), we obtain
\[
\text{TGV}_{\lambda_1, \lambda_2}(w_{\lambda_1, \lambda_2}) \leq \text{TGV}_{\lambda_1, \lambda_2}\left(\frac{\mu_2}{\|u^\lambda\|_{s, TV_1^2}} v_{\mu_1}^1\right) + \text{TGV}_{\lambda_1, \lambda_2}\left(v_{\mu_2}^2\right)
\]
\[
\leq \lambda_1 \frac{\mu_1}{\|u^\lambda\|_{s, TV_1^2}} TV^1(v_{\mu_1}^1) + \lambda_2 TV^2(v_{\mu_2}^2).
\]

Due to the definition of \(\text{TGV}_{\lambda_1, \lambda_2}\) and the choice of the parameters \(\lambda_1, \lambda_2\), we have that
\[w_{\lambda_1, \lambda_2} := w_{\lambda_1, \lambda_2} - u^\delta \in \mathcal{B}_{\text{TGV}_{\lambda_1, \lambda_2}},\]
such that
\[
- \int_{-1}^{1} w_{\lambda_1, \lambda_2}(w_{\lambda_1, \lambda_2} - u^\delta) \, dx \leq \text{TGV}_{\lambda_1, \lambda_2}(w_{\lambda_1, \lambda_2}).
\]

In order to simplify the left side, we calculate
\[
- \int_{-1}^{1} v_{\mu_1}^1(w_{\lambda_1, \lambda_2} - u^\delta) \, dx = -\frac{\mu_1}{\|u^\lambda\|_{s, TV_1^2}} \int_{-1}^{1} v_{\mu_1}^1(v_{\mu_1}^1 - u^\delta) \, dx
\]
\[
= \frac{\mu_1}{\|u^\lambda\|_{s, TV_1^2}} \mu_1 TV^1(v_{\mu_1}^1)
\]
and
\[
- \int_{-1}^{1} v_{\mu_2}^2(w_{\lambda_1, \lambda_2} - u^\delta) \, dx
\]
\[
= -\left(1 - \frac{\mu_2}{\|u^\lambda\|_{s, TV_1^2}}\right) \frac{\mu_2}{\|u^\lambda\|_{s, TV_1^2}} \int_{-1}^{1} u^\delta (v_{\mu_1}^1 - u^\delta) \, dx
\]
\[
= -\left(1 - \frac{\mu_2}{\|u^\lambda\|_{s, TV_1^2}}\right) \frac{\mu_2}{\|u^\lambda\|_{s, TV_1^2}} \left(\int_{-1}^{1} |x| (v_{\mu_1}^1 - u^\delta) \, dx - \frac{1}{2} \int_{-1}^{1} (v_{\mu_1}^1 - u^\delta) \, dx\right)
\]
\[
= -\left(1 - \frac{\mu_2}{\|u^\lambda\|_{s, TV_1^2}}\right) \frac{\mu_2}{\|u^\lambda\|_{s, TV_1^2}} \left(\int_{-1}^{0} \sigma^1[v_{\mu_1}^1] \, dx - \int_{0}^{1} \sigma^1[v_{\mu_1}^1] \, dx\right)
\]
\[
= \lambda_2 TV^2(v_{\mu_2}^2).
\]

In total we obtain
\[
- \int_{-1}^{1} w_{\lambda_1, \lambda_2} (w_{\lambda_1, \lambda_2} - u^\delta) \, dx
\]
\[
= \frac{\mu_1}{\|u^\lambda\|_{s, TV_1^2}} TV^1(v_{\mu_1}^1) + \lambda_2 TV^2(v_{\mu_2}^2).
\]
Comparing with (6.10) and (6.11) we have
\[ \text{TGV}_{\lambda_1, \lambda_2}(w_{\lambda_1, \lambda_2}) = -\int w_{\lambda_1, \lambda_2} (w_{\lambda_1, \lambda_2} - u^\delta), \]
which together with Lemma \ref{lem:5.7} implies that \( w_{\lambda_1, \lambda_2} \) is a minimizer of \( G \).

7. Example 2 Consider now as second test-data
\[ u^\delta(x) = 1_{[-\frac{1}{2}, \frac{1}{2}]}(x) - \frac{1}{2}, \quad (7.1) \]
where \( 1_{[a,b]} \) is the indicator function of the interval \( [a,b] \). Then
\[ \|u^\delta\|_{*,\text{TV}_1} = \frac{1}{4}, \quad \|u^\delta\|_{*,\text{TV}_2} = \frac{1}{8}. \]

First we calculate minimizers of \( F^i \), as defined in (2.1), in order to obtain the sets \( (\lambda_1, \lambda_2) \), where, according to Lemma \ref{lem:5.3} the TGV_{\lambda_1, \lambda_2}-minimizers are equal to some TV^i-minimizers.

7.1. \( F^1 \)-minimizers From \( \Pi \), we know that \( v^*_1 := \left(1 - \frac{\lambda_1}{\|u^\delta\|_{*,\text{TV}_1}}\right) u^\delta \) minimizes \( F^1 \) with the test data \( u^\delta \). Furthermore, we have
\[ \|v^*_1\|_{*,\text{TV}_1} = \lambda_1 \frac{\|u^\delta\|_{*,\text{TV}_2}}{\|u^\delta\|_{*,\text{TV}_1}}. \]
Thus, the coefficients $d_i$ keeping in mind that $x$ parameters, where the $G$-minimizer might be different to the $F^1$ or $F^2$-minimizer.

We write the solutions in the form

$$v_{\lambda_2}^2 = \begin{cases} u^\delta(x) & \text{for } |x| < x_1 \\ k_1 |x| + d_1 & \text{for } x_1 \leq |x| < x_2 \\ k_2 |x| + d_2 & \text{for } x_2 \leq |x| \leq 1 \end{cases} \tag{7.3}$$

keeping in mind that $x_1$ can be 0 (third case), or $x_2$ can be larger then one (third and second case). $v_{\lambda_1}^1$ is bending at $x_1$, such that $\sigma^2 [v_{\lambda_1}^1]$ is extremal at $x_0$ (hence $\pm \lambda_2$).

Thus, the coefficients $d_i, k_i$ are determined by the following equations:

$$\sigma^2 [v_{\lambda_1}^1] (x_1) = \lambda_2, \quad \sigma^1 [v_{\lambda_1}^1] (x_1) = 0 \quad \text{cases 1,2,3}$$

$$\sigma^2 [v_{\lambda_1}^1] (x_2) = -\lambda_2, \quad \sigma^1 [v_{\lambda_1}^1] (x_2) = 0 \quad \text{case 1}.$$

7.3. $G$-minimizers We consider the same approach as for the previous example. Hence, first we calculate the set $\Lambda$ as in Definition 5.4 which is illustrated as the green (solid) set in Figure 7.3. We have:

$$\partial \Lambda = \left\{ \left( \lambda_1, \| v_{\lambda_1}^1 \|_{*,TV_1^2} \right) : \lambda_1 \in \left[ 0, \| u^\delta \|_{*,TV_1^2} \right] \right\} \cup \left\{ \left( \| v_{\lambda_2}^2 \|_{*,TV_1^2}, \lambda_2 \right) : \lambda_2 \in \left[ 0, \| u^\delta \|_{*,TV_1^2} \right] \right\}.$$
Next we set up a general Ansatzfunction $w_{\lambda_1, \lambda_2}$ of piecewise affine functions, that is bending, once, twice or four times and jumping at $x = \pm 0.5$. Setting $w^*_{\lambda_1, \lambda_2} = w_{\lambda_1, \lambda_2} - u^\delta$, we find the coefficients (of the piecewise affine functions) by solving a number of non-linear equations coming from the conditions - $\sigma^1[w^*_{\lambda_1, \lambda_2}](x) = \lambda_1$ whenever the Ansatzfunction jumps and $\sigma^2[w^*_{\lambda_1, \lambda_2}](x) = \lambda_2, \sigma^1[w^*_{\lambda_1, \lambda_2}](x) = 0$, whenever the Ansatzfunction bends. We omit the explicit formulas and further calculations.

Then for $(\lambda_1, \lambda_2) \in \Lambda$ given, we found that the minimizers of $G$ can be written in a compact form:

**Theorem 7.1.** Let $(\lambda_1, \lambda_2) \in \Lambda$ and $\mu_1, \mu_2$ be such that

$$\lambda_2 = \frac{\mu_1 \mu_2}{\|u^\delta\|_{s, TV^1_1}}, \quad \lambda_1 = \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} \|v_{\mu_2}^{2s}\|_{s, TV^1_1};$$

Then for $u^\delta$ as in (7.1)

$$u_{\lambda_1, \lambda_2} = v_{\mu_1}^1 + \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} v_{\mu_2}^2$$

(7.4)

minimize $G$.

**Proof.** Lemma 4.1 states that $u_{\lambda_1, \lambda_2}$ is a minimizer if $TGV_{\lambda_1, \lambda_2}(u_{\lambda_1, \lambda_2}) = - \int_{-1}^1 u_{\lambda_1, \lambda_2}(u_{\lambda_1, \lambda_2} - u^\delta) \, dx$ (see (4.4)) and $u^*_{\lambda_1, \lambda_2} \in B^1_{TGV_{\lambda_1, \lambda_2}}$.

Using (7.4) we can estimate $TGV_{\lambda_1, \lambda_2}(u_{\lambda_1, \lambda_2})$ by

$$TGV_{\lambda_1, \lambda_2}(u_{\lambda_1, \lambda_2}) \leq TGV_{\lambda_1, \lambda_2}(v_{\mu_1}^1) + \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} TGV_{\lambda_1, \lambda_2}(v_{\mu_2}^2)$$

$$\leq TV_{\lambda_1}^1 (v_{\mu_1}^1) + \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} TV_{\lambda_2}^2 (v_{\mu_2}^2).$$

Since $v_{\mu_1}^1 = 1 - \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} u^\delta$, we have

$$u^*_{\lambda_1, \lambda_2} = \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} v_{\mu_2}^{2s}.$$ 

Note that from the choice of the parameters $\mu_1, \mu_2$ we have

$$\lambda_1 = \lambda_1 \|u^*_{\lambda_1, \lambda_2}\|_{s, TV^1_1}$$

$$= \left\| \sigma^1 \left[ v_{\mu_1}^1 + \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} v_{\mu_2}^2 - u^\delta \right] \right\|_{L^\infty}$$

$$= \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} \left\| \sigma^1 \left[ v_{\mu_2}^2 - u^\delta \right] \right\|_{L^\infty}$$

$$= \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} \lambda_1 \left\| v_{\mu_2}^{2s}\|_{s, TV^1_1};$$

and also

$$\lambda_2 = \lambda_2 \|u^*_{\lambda_1, \lambda_2}\|_{s, TV^2_2}$$

$$= \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} \left\| \sigma^2 \left[ v_{\mu_2}^2 - u^\delta \right] \right\|_{L^\infty}$$

$$= \frac{\mu_1}{\|u^\delta\|_{s, TV^1_1}} \lambda_2 \left\| v_{\mu_2}^{2s}\|_{s, TV^2_2};$$

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such that $u^*_{\lambda_1, \lambda_2} \in \mathcal{B}^*_{TGV_{\lambda_1, \lambda_2}}$ and

$$- \int_{-1}^{1} u_{\lambda_1, \lambda_2} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx \leq TGV_{\lambda_1, \lambda_2}(u_{\lambda_1, \lambda_2}) . \tag{7.5}$$

Moreover, we can write

$$\int_{-1}^{1} u_{\lambda_1, \lambda_2} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx = \int_{-1}^{1} v^1_{\mu_1} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx + \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}} \int_{-1}^{1} v^2_{\mu_2} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx.$$

Since $v^i_{\mu_i} = \left(1 - \frac{\mu_i}{\|u^\delta\|_{s,TV_1^i}}\right) u^\delta$, and $u^\delta(x) \in \{-\frac{1}{2}, \frac{1}{2}\}$ we obtain

$$\int_{-1}^{1} v^1_{\mu_1} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx = \frac{1}{2} \left(1 - \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}}\right) \left( -\int_{-1}^{\frac{1}{2}} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx + \int_{\frac{1}{2}}^{1} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx - \int_{-1}^{1} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx \right)
= \frac{1}{2} \left(1 - \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}}\right) \left(-\sigma^1[u^*_{\lambda_1, \lambda_2}] (-\frac{1}{2}) + \sigma^1[u^*_{\lambda_1, \lambda_2}] (\frac{1}{2}) - \sigma^1[u^*_{\lambda_1, \lambda_2}] (-\frac{1}{2}) \right)$$
$$-\sigma^1[u^*_{\lambda_1, \lambda_2}] (1) + \sigma^1[u^*_{\lambda_1, \lambda_2}] (\frac{1}{2}) \right) .$$

Now by the choice of the parameter $\lambda_1$ we have $\sigma^1[u^*_{\lambda_1, \lambda_2}] (-\frac{1}{2}) = +\lambda_1$ and $\sigma^1[u^*_{\lambda_1, \lambda_2}] (\frac{1}{2}) = -\lambda_1$ such that the equation above simplifies to

$$- \int_{-1}^{1} v^1_{\mu_1} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx = \lambda_1 \left(1 - \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}}\right) = \lambda_1 TV^1 \left(v^1_{\mu_1}\right) .$$

Next it remains to calculate $\int_{-1}^{1} v^2_{\mu_2} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx$. Since

$$u_{\lambda_1, \lambda_2} - u^\delta = v^1_{\mu_1} - \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}} v^2_{\mu_2} - u^\delta = \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}} (v^2_{\mu_2} - u^\delta) ,$$

we have

$$\int_{-1}^{1} v^2_{\mu_2} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx = \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}} \int_{-1}^{1} v^2_{\mu_2} (v^2_{\mu_2} - u^\delta) \, dx$$
$$= - \frac{\mu_1}{\|u^\delta\|_{s,TV_1^i}} \mu_2 TV^2 \left(v^2_{\mu_2}\right) ,$$

where we used $TV^2 \left(v^2_{\mu_2}\right) = -\mu_2 \int_{-1}^{1} v^2_{\mu_2} (v^2_{\mu_2} - u^\delta) \, dx$, the optimality condition for $\mathcal{F}^1$-minimizers as in [4,3]. Hence in total, using the connections between $\lambda_i$ and $\mu_i$, 28
Fig. 7.2. \( u_{\lambda_1, \lambda_2} \) for fixed \( \lambda_1 \) and changing \( \lambda_2 \). In this particular case we have \((\lambda_1, \tilde{\lambda}_2), (\lambda_1, \lambda_2) \not\in \Lambda \) with \( \lambda_2 < \tilde{\lambda}_2 < \lambda_2 \), such that \( u_{\lambda_1, \tilde{\lambda}_2} = v_{\lambda_1}^1 \) and \( u_{\lambda_1, \lambda_2} = v_{\lambda_2}^2 \).

we obtain

\[
- \int_{-1}^{1} u_{\lambda_1, \lambda_2} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx = \lambda_1 \TV^1_1 (v_{\mu_1}^1) + \lambda_2 \mu_2 \TV^2_1 (v_{\mu_2}^2) .
\]

A Comparison with (7.5) shows that TGV_{\lambda_1, \lambda_2} (u_{\lambda_1, \lambda_2}) = - \int_{-1}^{1} u_{\lambda_1, \lambda_2} (u_{\lambda_1, \lambda_2} - u^\delta) \, dx, hence according to Lemma 4.1 \( u_{\lambda_1, \lambda_2} \) minimizes \( G \).

8. Example III Finally we consider \( u^\delta = x^2 - \frac{1}{3} \) but only sketch the different minimizers of \( F \) and \( G \) in order to show that in general, minimizers of \( G \) cannot be written as a sum of \( F \)-minimizers. We have \( \|u^\delta\|_{s, \TV^2_1} = \frac{1}{12} = 0.833, \|u^\delta\|_{s, \TV^1_1} = \frac{2\sqrt{3}}{27} \approx 0.12 \).

8.1. \( F^1 \)-minimizers. Since \( u^\delta \) is continuous, also the \( F^1 \)-minimizer is continuous. From the characterization of \( F_1 \) minimizers we know that \( v_{\lambda_1}^1 \) is either equal to \( u^\delta \) in an interval \((\pm c_1, \pm c_2)\) or constant \( u^\delta(c_1), u^\delta(c_2) \) in the other intervals. In Figure 8.1 (left), we illustrate \( v_{\lambda_1}^1 \) for different values of \( \lambda_1 \).

8.2. \( F^2 \)-minimizers. In this case, we have to consider two different types of minimizers.

- \( \lambda_2 \) large: (that is \( \lambda \sim \|u^\delta\|_{s, \TV^2_2} \)), \( v_{\lambda_2}^2 \) is piecewise constant and bending at \( x = 0 \). Such solutions are constructed by considering the Ansatzfunctions \( w(x, k) := k \left(|x| - \frac{1}{2}\right) \). The parameter \( k \) is determined such that \( |\sigma^2 [w(\cdot, k) - u^\delta](0)] = \lambda_2 \) (the Ansatzfunction \( w \) is bending at \( x = 0 \), hence the \( \sigma^2 \) of the dual minimizer has to be extremal, hence equal to \( \lambda_2 \)). This Ansatzfunction works until for some \( \lambda_2 = \tilde{\lambda}_2 \), we have \( u^\delta(0) = w(0, k) \).
Then for $\lambda_2 \leq \bar{\lambda}_2$, we use a different Ansatz function $w$ that satisfies: $w(x) = u^\delta(x)$ for $x \in (-c, c)$ and some $c > 0$ and $w(x)$ is affine linear in $(-1, -c) \cup (c, 1)$ and continuous at $x = \pm c$. The coefficients are determined such that $w \in H^2$ and $|\sigma^2 [w - u^\delta](c)| = \lambda_2$.

We illustrated both types of solutions in Figure 8.1 (right).

8.3. $G$-minimizers. For $(\lambda_1, \lambda_2) \in \Lambda$ as in Definition 5.4, we set up an Ansatz function that satisfies the following:

- $w$ is continuous,
- $w(x) = w(-x)$,
- $w(x) = u^\delta(x)$ for $x \in (c_2, c_3)$ and $0 \leq c_2 < c_3 \leq 1$,
- either $w$ is bending at $x = 0$, or $w(x) = u^\delta(x)$ for $x \in [0, c_1)$ with $c_1 < c_2$,
- $w$ is piecewise affine linear else.

We illustrate minimizers for different choices of $(\lambda_1, \lambda_2)$ in Figure 8.1.

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Fig. 8.1. Minimizers of $F_1$ (left), $G$ (middle), $F_2$ (right), for different parameters.

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