Dual variational formulations for a large class of non-convex models in the calculus of variations

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Abstract

This article develops dual variational formulations for a large class of models in variational optimization. The results are established through basic tools of functional analysis, convex analysis and duality theory. The main duality principle is developed as an application to a Ginzburg-Landau type system in superconductivity in the absence of a magnetic field. In the first part final sections, we develop new general dual convex variational formulations, more specifically, dual formulations with a large region of convexity around the critical points which are suitable for the non-convex optimization for a large class of models in physics and engineering. Finally, in the last section we present some numerical results concerning the generalized method of lines applied to a Ginzburg-Landau type equation.

1 Introduction

In this section we establish a dual formulation for a large class of models in non-convex optimization.

The main duality principle is applied to the Ginzburg-Landau system in superconductivity in an absence of a magnetic field.

Such results are based on the works of J.J. Telega and W.R. Bielski [2, 3, 13, 14] and on a D.C. optimization approach developed in Toland [15].

At this point we start to describe the primal and dual variational formulations.

Let $\Omega \subset \mathbb{R}^3$ be an open, bounded, connected set with a regular (Lipschitzian) boundary denoted by $\partial\Omega$.

For the primal formulation we consider the functional $J:U\to\mathbb{R}$ where

$$J(u) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx + \frac{\alpha}{2} \int_{\Omega} (u^2 - \beta)^2 \, dx - \langle u, f \rangle_{L^2}.$$
 (1)

Here we assume $\alpha>0,\beta>0,\gamma>0,$ $U=W_0^{1,2}(\Omega),$ $f\in L^2(\Omega).$ Moreover we denote $Y=Y^*=L^2(\Omega).$

Define also $G_1: U \to \mathbb{R}$ by

$$G_1(u) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx,$$

 $G_2: U \times Y \to \mathbb{R}$ by

$$G_2(u,v) = \frac{\alpha}{2} \int_{\Omega} (u^2 - \beta + v)^2 dx + \frac{K}{2} \int_{\Omega} u^2 dx,$$

and $F: U \to \mathbb{R}$ by

$$F(u) = \frac{K}{2} \int_{\Omega} u^2 \, dx,$$

where $K \gg \gamma$.

It is worth highlighting that in such a case

$$J(u) = G_1(u) + G_2(u, 0) - F(u) - \langle u, f \rangle_{L^2}, \ \forall u \in U.$$

Furthermore, define the following specific polar functionals specified, namely, $G_1^*:[Y^*]^2\to\mathbb{R}$ by

$$G_1^*(v_1^* + z^*) = \sup_{u \in U} \{ \langle u, v_1^* + z^* \rangle_{L^2} - G_1(u) \}$$
$$= \frac{1}{2} \int_{\Omega} [(-\gamma \nabla^2)^{-1} (v_1^* + z^*)] (v_1^* + z^*) dx, \tag{2}$$

 $G_2^*: [Y^*]^2 \to \mathbb{R}$ by

$$G_1^*(v_2^*, v_0^*) = \sup_{(u,v) \in U \times Y} \{ \langle u, v_2^* \rangle_{L^2} + \langle v, v_0^* \rangle_{L^2} - G_2(u, v) \}$$

$$= \frac{1}{2} \int_{\Omega} \frac{(v_2^*)^2}{2v_0^* + K} dx$$

$$+ \frac{1}{2\alpha} \int_{\Omega} (v_0^*)^2 dx + \beta \int_{\Omega} v_0^* dx, \qquad (3)$$

if $v_0^* \in B^*$ where

$$B^* = \{v_0^* \in Y^* : 2v_0^* + K > K/2 \text{ in } \Omega\},\$$

and finally, $F^*: Y^* \to \mathbb{R}$ by

$$F^*(z^*) = \sup_{u \in U} \{ \langle u, z^* \rangle_{L^2} - F(u) \}$$
$$= \frac{1}{2K} \int_{\Omega} (z^*)^2 dx. \tag{4}$$

Define also

$$A^* = \{v^* = (v_1^*, v_2^*, v_0^*) \in [Y^*]^2 \times B^* : v_1^* + v_2^* - f = 0, \text{ in } \Omega\},\$$

 $J^*: [Y^*]^4 \to \mathbb{R}$ by

$$J^*(v^*, z^*) = -G_1^*(v_1^* + z^*) - G_2^*(v_2^*, v_0^*) + F^*(z^*)$$

and $J_1^*: [Y^*]^4 \times U \to \mathbb{R}$ by

$$J_1^*(v^*, z^*, u) = J^*(v^*, z^*) + \langle u, v_1^* + v_2^* - f \rangle_{L^2}.$$

2 The main duality principle, a convex dual formulation and the concerning proximal primal functional

Our main result is summarized by the following theorem.

Theorem 2.1. Considering the definitions and statements in the last section, suppose also $(\hat{v}^*, \hat{z}^*, u_0) \in [Y^*]^2 \times B^* \times Y^* \times U$ is such that

$$\delta J_1^*(\hat{v}^*, \hat{z}^*, u_0) = \mathbf{0}.$$

Under such hypotheses, we have

$$\delta J(u_0) = \mathbf{0},$$
$$\hat{v}^* \in A^*$$

and

$$J(u_0) = \inf_{u \in U} \left\{ J(u) + \frac{K}{2} \int_{\Omega} |u - u_0|^2 dx \right\}$$

= $J^*(\hat{v}^*, \hat{z}^*)$
= $\sup_{v^* \in A^*} \left\{ J^*(v^*, \hat{z}^*) \right\}.$ (5)

Proof. Since

$$\delta J_1^*(\hat{v}^*, \hat{z}^*, u_0) = \mathbf{0}$$

from the variation in v_1^* we obtain

$$-\frac{(\hat{v}_1^* + \hat{z}^*)}{-\gamma \nabla^2} + u_0 = 0 \text{ in } \Omega,$$

so that

$$\hat{v}_1^* + \hat{z}^* = -\gamma \nabla^2 u_0.$$

From the variation in v_2^* we obtain

$$-\frac{\hat{v}_2^*}{2\hat{v}_0^* + K} + u_0 = 0, \text{ in } \Omega.$$

From the variation in v_0^* we also obtain

$$\frac{(\hat{v}_2^*)^2}{(2\hat{v}_0^* + K)^2} - \frac{\hat{v}_0^*}{\alpha} - \beta = 0$$

and therefore,

$$\hat{v}_0^* = \alpha(u_0^2 - \beta).$$

From the variation in u we get

$$\hat{v}_1^* + \hat{v}_2^* - f = 0$$
, in Ω

and thus

$$\hat{v}^* \in A^*$$
.

Finally, from the variation in z^* , we obtain

$$-\frac{(\hat{v}_{1}^{*}+\hat{z}^{*})}{-\gamma\nabla^{2}}+\frac{\hat{z}^{*}}{K}=0, \text{ in } \Omega.$$

so that

$$-u_0 + \frac{\hat{z}^*}{K} = 0,$$

that is,

$$\hat{z}^* = Ku_0 \text{ in } \Omega.$$

From such results and $\hat{v}^* \in A^*$ we get

$$0 = \hat{v}_1^* + \hat{v}_2^* - f$$

= $-\gamma \nabla^2 u_0 - \hat{z}^* + 2(v_0^*)u_0 + Ku_0 - f$
= $-\gamma \nabla^2 u_0 + 2\alpha(u_0^2 - \beta)u_0 - f$, (6)

so that

$$\delta J(u_0) = \mathbf{0}.$$

Also from this and from the Legendre transform proprieties we have

$$G_1^*(\hat{v}_1^* + \hat{z}^*) = \langle u_0, \hat{v}_1^* + \hat{z}^* \rangle_{L^2} - G_1(u_0),$$

$$G_2^*(\hat{v}_2^*, \hat{v}_0^*) = \langle u_0, \hat{v}_2^* \rangle_{L^2} + \langle 0, v_0^* \rangle_{L^2} - G_2(u_0, 0),$$

$$F^*(\hat{z}^*) = \langle u_0, \hat{z}^* \rangle_{L^2} - F(u_0)$$

and thus we obtain

$$J^{*}(\hat{v}^{*}, \hat{z}^{*}) = -G_{1}^{*}(\hat{v}_{1}^{*} + \hat{z}^{*}) - G_{2}^{*}(\hat{v}_{2}^{*}, \hat{v}_{0}^{*}) + F^{*}(\hat{z}^{*})$$

$$= -\langle u_{0}, \hat{v}_{1}^{*} + \hat{v}_{2}^{*} \rangle + G_{1}(u_{0}) + G_{2}(u_{0}, 0) - F(u_{0})$$

$$= -\langle u_{0}, f \rangle_{L^{2}} + G_{1}(u_{0}) + G_{2}(u_{0}, 0) - F(u_{0})$$

$$= J(u_{0}).$$
(7)

Summarizing, we have got

$$J^*(\hat{v}^*, \hat{z}^*) = J(u_0). \tag{8}$$

On the other hand

$$J^{*}(\hat{v}^{*}, \hat{z}^{*}) = -G_{1}^{*}(\hat{v}_{1}^{*} + \hat{z}^{*}) - G_{2}^{*}(\hat{v}_{2}^{*}, \hat{v}_{0}^{*}) + F^{*}(\hat{z}^{*})$$

$$\leq -\langle u, \hat{v}_{1}^{*} + \hat{z}^{*} \rangle_{L^{2}} - \langle u, \hat{v}_{2}^{*} \rangle_{L^{2}} - \langle 0, v_{0}^{*} \rangle_{L^{2}} + G_{1}(u) + G_{2}(u, 0) + F^{*}(\hat{z}^{*})$$

$$= -\langle u, f \rangle_{L^{2}} + G_{1}(u) + G_{2}(u, 0) - \langle u, \hat{z}^{*} \rangle_{L^{2}} + F^{*}(\hat{z}^{*})$$

$$= -\langle u, f \rangle_{L^{2}} + G_{1}(u) + G_{2}(u, 0) - F(u) + F(u) - \langle u, \hat{z}^{*} \rangle_{L^{2}} + F^{*}(\hat{z}^{*})$$

$$= J(u) + \frac{K}{2} \int_{\Omega} u^{2} dx - \langle u, \hat{z}^{*} \rangle_{L^{2}} + F^{*}(\hat{z}^{*})$$

$$= J(u) + \frac{K}{2} \int_{\Omega} u^{2} dx - K\langle u, u_{0} \rangle_{L^{2}} + \frac{K}{2} \int_{\Omega} u_{0}^{2} dx$$

$$= J(u) + \frac{K}{2} \int_{\Omega} |u - u_{0}|^{2} dx, \ \forall u \in U.$$

$$(9)$$

Finally by a simple computation we may obtain the Hessian

$$\left\{\frac{\partial^2 J^*(v^*, z^*)}{\partial (v^*)^2}\right\} < \mathbf{0}$$

in $[Y^*]^2 \times B^* \times Y^*$, so that we may infer that J^* is concave in v^* in $[Y^*]^2 \times B^* \times Y^*$. Therefore, from this, (8) and (9), we have

$$J(u_0) = \inf_{u \in U} \left\{ J(u) + \frac{K}{2} \int_{\Omega} |u - u_0|^2 dx \right\}$$

= $J^*(\hat{v}^*, \hat{z}^*)$
= $\sup_{v^* \in A^*} \left\{ J^*(v^*, \hat{z}^*) \right\}.$ (10)

The proof is complete.

3 A primal dual variational formulation

In this section we develop a more general primal dual variational formulation suitable for a large class of models in non-convex optimization.

Consider again $U = W_0^{1,2}(\Omega)$ and let $G: U \to \mathbb{R}$ and $F: U \to \mathbb{R}$ be three times Fréchet differentiable functionals. Let $J: U \to \mathbb{R}$ be defined by

$$J(u) = G(u) - F(u), \ \forall u \in U.$$

Assume $u_0 \in U$ is such that

$$\delta J(u_0) = \mathbf{0}$$

and

$$\delta^2 J(u_0) > \mathbf{0}.$$

Denoting $v^* = (v_1^*, v_2^*)$, define $J^* : U \times Y^* \times Y^* \to \mathbb{R}$ by

$$J^*(u, v^*) = \frac{1}{2} \|v_1^* - G'(u)\|_2^2 + \frac{1}{2} \|v_2^* - F'(u)\|_2^2 + \frac{1}{2} \|v_1^* - v_2^*\|_2^2$$
(11)

Denoting $L_1^*(u, v^*) = v_1^* - G'(u)$ and $L_2^*(u, v^*) = v_2^* - F'(u)$, define also

$$C^* = \left\{ (u, v^*) \in U \times Y^* \times Y^* : \|L_1^*(u, v_1^*)\|_{\infty} \le \frac{1}{K} \text{ and } \|L_2^*(u, v_1^*)\|_{\infty} \le \frac{1}{K} \right\},$$

for an appropriate K > 0 to be specified.

Observe that in C^* the Hessian of J^* is given by

$$\{\delta^2 J^*(u, v^*)\} = \left\{ \begin{array}{ll} G''(u)^2 + F''(u)^2 + \mathcal{O}(1/K) & -G''(u) & -F''(u) \\ -G''(u) & 2 & -1 \\ -F''(u) & -1 & 2 \end{array} \right\},\tag{12}$$

Observe also that

$$\det\left\{\frac{\partial^2 J^*(u,v^*)}{\partial v_1^* \partial v_2^*}\right\} = 3,$$

and

$$\det\{\delta^2 J^*(u, v^*)\} = (G''(u) - F''(u))^2 + \mathcal{O}(1/K) = (\delta^2 J(u))^2 + \mathcal{O}(1/K).$$

Define now

$$\hat{v}_1^* = G'(u_0),$$

$$\hat{v}_2^* = F'(u_0),$$

so that

$$\hat{v}_1^* - \hat{v}_2^* = \mathbf{0}.$$

From this we may infer that $(u_0, \hat{v}_1^*, \hat{v}_2^*) \in C^*$ and

$$J^*(u_0, \hat{v}^*) = 0 = \min_{(u, v^*) \in C^*} J^*(u, v^*).$$

Moreover, for K > 0 sufficiently big, J^* is convex in a neighborhood of (u_0, \hat{v}^*) . Therefore, in the last lines, we have proven the following theorem.

Theorem 3.1. Under the statements and definitions of the last lines, there exist $r_0 > 0$ and $r_1 > 0$ such that

$$J(u_0) = \min_{u \in B_{r_0}(u_0)} J(u)$$

and $(u_0, \hat{v}_1^*, \hat{v}_2^*) \in C^*$ is such that

$$J^*(u_0, \hat{v}^*) = 0 = \min_{(u, v^*) \in U \times [Y^*]^2} J^*(u, v^*).$$

Moreover, J^* is convex in

$$B_{r_1}(u_0, \hat{v}^*).$$

4 One more primal dual variational formulation for variational optimization

Our next result is a new primal dual variational formulation.

Consider again the functional, as defined in the previous sections, $J:U\to\mathbb{R}$ where

$$J(u) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx + \frac{\alpha}{2} \int_{\Omega} (u^2 - \beta)^2 \, dx - \langle u, f \rangle_{L^2},$$
(13)

where $\alpha>0,\ \beta>0,\ \gamma>0,\ U=W_0^{1,2}(\Omega)\ Y=Y^*=L^2(\Omega)$ and $f\in L^\infty(\Omega).$

Define

$$B^+ = \{ u \in U : uf \ge 0, \text{ in } \Omega \}$$

and

$$E^+ = \{ u \in U : \delta^2 J(u) > \mathbf{0} \}.$$

Observe that

$$J(u) = J(u) - \frac{K}{2} \int_{\Omega} u^{4} dx + \langle v_{0}^{*}, u^{2} \rangle_{L^{2}}$$

$$-\langle v_{0}^{*}, u^{2} \rangle_{L^{2}} + \frac{K}{2} \int_{\Omega} u^{4} dx$$

$$\geq J(u) - \frac{K}{2} \int_{\Omega} u^{4} dx + \langle v_{0}^{*}, u^{2} \rangle_{L^{2}}$$

$$\inf_{u \in U} \left\{ -\langle v_{0}^{*}, u^{2} \rangle_{L^{2}} + \frac{K}{2} \int_{\Omega} u^{4} dx \right\}$$

$$= J(u) - \frac{K}{2} \int_{\Omega} u^{4} dx + \langle v_{0}^{*}, u^{2} \rangle_{L^{2}} - \frac{1}{2K} \int_{\Omega} (v_{0}^{*})^{2} dx.$$

$$(14)$$

For a critical point we have the relation

$$v_0^* = Ku^2.$$

With such a relation in mind, we define the following approximate primal dual functional

$$J^*(u, v_0^*) = J(u) + \frac{K}{2} \int_{\Omega} u^4 dx - \frac{1}{2K} \int_{\Omega} (v_0^*)^2 dx + \frac{K_1}{2} \int_{\Omega} (v_0^* - Ku^2)^2 dx,$$

where $K_1 \gg K \gg 1$.

Observe also that

$$\det\{\delta^2 J^*(u, v_0^*)\} = -\frac{\delta^2 J(u)}{K} + K_1 \delta^2 J(u) - 6u^2 + 2K_1 v_0^* + 2K^2 K_1^2 u^2 - 2K K_1^2 v_0^*.$$
(15)

At this point, we define

$$A_1^+ = \{(u, v_0^*) \in U \times Y^* : Ku^2 - v_0^* \le 0, \text{ in } \Omega\}$$

and

$$A_2^+ = \{(u, v_0^*) \in U \times Y^* : Ku^2 - v_0^* = 0, \text{ in } \Omega\}.$$

Define also $A^+ = A_2^+ \cap C^+$, where

$$C^+ = \{(u, v_0^*) \in U \times Y^* : u \in B^+ \cap E^+\}.$$

Thus, we have

$$\det\{\delta^2 J^*(u, v_0^*)\} = \mathcal{O}(K_1) > \mathbf{0}$$

on A + .

Consider the following convex primal dual problem:

Locally minimize $J^*(u, v_0^*)$ on A^+ .

Define $\hat{J}^*: U \times Y^* \times Y^* \to \mathbb{R}$, by

$$\hat{J}^*(u, v_0^*, \lambda) = J^*(u, v_0^*) + \int_{\Omega} \lambda (Ku^2 - v_0^*) \, dx,$$

where $\lambda \in Y^*$ is an appropriate Lagrange multiplier.

Considering such statements and definitions, our main result in this section is summarized by the following theorem.

Theorem 4.1. Let $(u_0, \hat{v}_0^*) \in A^+$ and $\lambda_0 \in Y^*$ be such that

$$\delta \hat{J}^*(u_0, \hat{v}_0^*, \lambda_0) = \mathbf{0}.$$

Under such hypotheses,

$$\delta J(u_0) = \mathbf{0},$$

and there exists r > 0 such that

$$\hat{J}^*(u_0,\hat{v}_0^*,\lambda_0) = J^*(u_0,\hat{v}_0^*) = \min_{(u,v_0^*) \in A^+ \cap B_r(u_0,\hat{v}_0^*)} J^*(u,v_0^*) = \min_{u \in B^+ \cap E^+} J(u) = J(u_0).$$

Proof. Since $(u_0, v_0^*) \in A_2^+$ and $u_0 \in B^+ \cap E^+$, we have that there exists r > 0 such that

$$\det\{\delta^2 J^*(u, v_0^*)\} = \mathcal{O}(K_1) > \mathbf{0}$$

in $B_r(u_0, \hat{v}_0^*)$.

From this, $\delta \hat{J}^*(u_0, \hat{v}_0^*, \lambda_0) = \mathbf{0}$, the other hypotheses and the expression for λ_0 indicated in the next lines, we have that

$$J^*(u_0, \hat{v}_0^*) = \min_{(u, v_0^*) \in A^+ \cap B_r(u_0, \hat{v}_0^*)} J^*(u, v_0^*).$$

From the variation of \hat{J}^* in u, we get

$$\delta J(u_0) + 2Ku_0^3 + K_1(\hat{v}_0^* - Ku_0^2)(-2Ku_0) + 2K\lambda_0 u_0 = \mathbf{0}.$$
(16)

From the variation of \hat{J}^* in v_0^* , we have

$$-\frac{\hat{v}_0^*}{K} + K_1(\hat{v}_0^* - Ku_0^2) - \lambda_0 = 0,$$

Finally, from the variation of \hat{J}^* in λ , we have

$$\hat{v}_0^* - K u_0^2 = 0$$

Replacing such results we have obtained

$$\lambda_0 = -u_0$$

so that replacing these last results into (16), we get

$$\delta J(u_0) = \mathbf{0}.$$

Also from

$$\hat{v}_0^* - Ku_0^2 = 0,$$

we have

$$\hat{J}^*(u_0, \hat{v}_0^*, \lambda_0) = J^*(u_0, \hat{v}_0^*)
= J(u_0).$$
(17)

Finally, observe that from similar results in [5], we may infer that J is convex on the convex set

$$B^+ \cap E^+$$
.

Joining the pieces, considering that $\delta J(u_0) = \mathbf{0}$, we have got

$$\hat{J}^*(u_0, \hat{v}_0^*, \lambda_0) = J^*(u_0, \hat{v}_0^*) = \min_{(u, v_0^*) \in A^+ \cap B_r(u_0, \hat{v}_0^*)} J^*(u, v_0^*) = \min_{u \in B^+ \cap E^+} J(u) = J(u_0).$$

The proof is complete.

5 One more duality principle and a concerning primal dual variational formulation

In this section we establish a new duality principle and a related primal dual formulation. The results are based on the approach of Toland, [15].

5.1 Introduction

Let $\Omega \subset \mathbb{R}^3$ be an open, bounded, connected set with a regular (Lipschitzian) boundary denoted by $\partial\Omega$.

Let $J: V \to \mathbb{R}$ be a functional such that

$$J(u) = G(u) - F(u), \forall u \in V,$$

where $V = W_0^{1,2}(\Omega)$.

Suppose G, F are both three times Fréchet differentiable convex functionals such that

$$\frac{\partial^2 G(u)}{\partial u^2} > 0$$

and

$$\frac{\partial^2 F(u)}{\partial u^2} > 0$$

 $\forall u \in V$.

Assume also there exists $\alpha_1 \in \mathbb{R}$ such that

$$\alpha_1 = \inf_{u \in V} J(u).$$

Moreover, suppose that if $\{u_n\} \subset V$ is such that

$$||u_n||_V \to \infty$$

then

$$J(u_n) \to +\infty$$
, as $n \to \infty$.

At this point we define $J^{**}: V \to \mathbb{R}$ by

$$J^{**}(u) = \sup_{(v^*, \alpha) \in H^*} \{ \langle u, v^* \rangle + \alpha \},$$

where

$$H^* = \{ (v^*, \alpha) \in V^* \times \mathbb{R} : \langle v, v^* \rangle_V + \alpha \le F(v), \ \forall v \in V \}.$$

Observe that $(0, \alpha_1) \in H^*$, so that

$$J^{**}(u) \ge \alpha_1 = \inf_{u \in V} J(u).$$

On the other hand, clearly we have

$$J^{**}(u) \le J(u), \ \forall u \in V,$$

so that we have got

$$\alpha_1 = \inf_{u \in V} J(u) = \inf_{u \in V} J^{**}(u).$$

Let $u \in V$.

Since J is strongly continuous, there exist $\delta > 0$ and A > 0 such that,

$$\alpha_1 \le J^{**}(v) \le J(v) \le A, \forall v \in B_{\delta}(u).$$

From this, considering that J^{**} is convex on V, we may infer that J^{**} is continuous at u, $\forall u \in V$

Hence J^{**} is strongly lower semi-continuous on V, and since J^{**} is convex we may infer that J^{**} is weakly lower semi-continuous on V.

Let $\{u_n\} \subset V$ be a sequence such that

$$\alpha_1 \le J(u_n) < \alpha_1 + \frac{1}{n}, \ \forall n \in \mathbb{N}.$$

Hence

$$\alpha_1 = \lim_{n \to \infty} J(u_n) = \inf_{u \in V} J(u) = \inf_{u \in V} J^{**}(u).$$

Suppose there exists a subsequence $\{u_{n_k}\}$ of $\{u_n\}$ such that

$$||u_{n_k}||_V \to \infty$$
, as $k \to \infty$.

From the hypothesis we have

$$J(u_{n_k}) \to +\infty$$
, as $k \to \infty$,

which contradicts

$$\alpha_1 \in \mathbb{R}$$
.

Therefore there exists K > 0 such that

$$||u_n||_V \le K, \ \forall u \in V.$$

Since V is reflexive, from this and the Katutani Theorem, there exists a subsequence $\{u_{n_k}\}$ of $\{u_n\}$ and $u_0 \in V$ such that

$$u_{n_k} \rightharpoonup u_0$$
, weakly in V .

Consequently, from this and considering that J^{**} is weakly lower semi-continuous, we have got

$$\alpha_1 = \liminf_{k \to \infty} J^{**}(u_{n_k}) \ge J^{**}(u_0),$$

so that

$$J^{**}(u_0) = \min_{u \in V} J^{**}(u).$$

Define $G^*, F^*: V^* \to \mathbb{R}$ by

$$G^*(v^*) = \sup_{u \in V} \{ \langle u, v^* \rangle_V - G(u) \},$$

and

$$F^*(v^*) = \sup_{u \in V} \{\langle u, v^* \rangle_V - F(u)\}.$$

Defining also $J^*: V \to \mathbb{R}$ by

$$J^*(v^*) = F^*(v^*) - G^*(v^*),$$

from the results in [15], we may obtain

$$\inf_{u \in V} J(u) = \inf_{v^* \in V^*} J^*(v^*),$$

so that

$$J^{**}(u_0) = \inf_{u \in V} J^{**}(u)$$

= $\inf_{u \in V} J(u) = \inf_{v^* \in V^*} J^*(v^*).$ (18)

Suppose now there exists $\hat{u} \in V$ such that

$$J(\hat{u}) = \inf_{u \in V} J(u).$$

From the standard necessary conditions, we have

$$\delta J(\hat{u}) = \mathbf{0},$$

so that

$$\frac{\partial G(\hat{u})}{\partial u} - \frac{\partial F(\hat{u})}{\partial u} = \mathbf{0}.$$

Define now

$$v_0^* = \frac{\partial F(\hat{u})}{\partial u}.$$

From these last two equations we obtain

$$v_0^* = \frac{\partial G(\hat{u})}{\partial u}.$$

From such results and the Legendre transform properties, we have

$$\hat{u} = \frac{\partial F^*(v_0^*)}{\partial v^*},$$

$$\hat{u} = \frac{\partial G^*(v_0^*)}{\partial v^*},$$

so that

$$\delta J^{*}(v_{0}^{*}) = \frac{\partial F^{*}(v_{0}^{*})}{\partial v^{*}} - \frac{\partial G^{*}(v_{0}^{*})}{\partial v^{*}} = \hat{u} - \hat{u} = \mathbf{0},$$

$$G^*(v_0^*) = \langle \hat{u}, v_0^* \rangle_V - G(\hat{u})$$

and

$$F^*(v_0^*) = \langle \hat{u}, v_0^* \rangle_V - F(\hat{u})$$

so that

$$\inf_{u \in V} J(u) = J(\hat{u})
= G(\hat{u}) - F(\hat{u})
= \inf_{v^* \in V^*} J^*(v^*)
= F^*(v_0^*) - G^*(v_0^*)
= J^*(v_0^*).$$
(19)

5.2 The main duality principle and a related primal dual variational formulation

Considering these last statements and results, we may prove the following theorem.

Theorem 5.1. Let $\Omega \subset \mathbb{R}^3$ be an open, bounded, connected set with a regular (Lipschitzian) boundary denoted by $\partial\Omega$.

Let $J: V \to \mathbb{R}$ be a functional such that

$$J(u) = G(u) - F(u), \forall u \in V,$$

where $V = W_0^{1,2}(\Omega)$.

Suppose G, F are both three times Fréchet differentiable functionals such that there exists K > 0 such that

$$\frac{\partial^2 G(u)}{\partial u^2} + K > 0$$

and

$$\frac{\partial^2 F(u)}{\partial u^2} + K > 0$$

 $\forall u \in V$.

Assume also there exists $u_0 \in V$ and $\alpha_1 \in \mathbb{R}$ such that

$$\alpha_1 = \inf_{u \in V} J(u) = J(u_0).$$

Assume $K_3 > 0$ is such that

$$||u_0||_{\infty} < K_3.$$

Define

$$\tilde{V} = \{ u \in V : ||u||_{\infty} \le K_3 \}.$$

Assume $K_1 > 0$ is such that if $u \in \tilde{V}$ then

$$\max \left\{ \|F'(u)\|_{\infty}, \|G'(u)\|_{\infty}, \|F''(u)\|_{\infty}, \|F'''(u)\|_{\infty}, \|G'''(u)\|_{\infty}, \|G'''(u)\|_{\infty} \right\} \leq K_{1}.$$

Suppose also

$$K \gg \max\{K_1, K_3\}.$$

Define $F_K, G_K : V \to \mathbb{R}$ by

$$F_K(u) = F(u) + \frac{K}{2} \int_{\Omega} u^2 dx,$$

and

$$G_K(u) = G(u) + \frac{K}{2} \int_{\Omega} u^2 dx,$$

 $\forall u \in V$.

Define also $G_K^*, F_K^*: V^* \to \mathbb{R}$ by

$$G_K^*(v^*) = \sup_{u \in V} \{ \langle u, v^* \rangle_V - G_K(u) \},$$

and

$$F_K^*(v^*) = \sup_{u \in V} \{ \langle u, v^* \rangle_V - F_K(u) \}.$$

Observe that since $u_0 \in V$ is such that

$$J(u_0) = \inf_{u \in V} J(u),$$

we have

$$\delta J(u_0) = \mathbf{0}.$$

Let $\varepsilon > 0$ be a small constant.

Define

$$v_0^* = \frac{\partial F_K(u_0)}{\partial u} \in V^*.$$

Under such hypotheses, defining $J_1^*: V \times V^* \to \mathbb{R}$ by

$$J_{1}^{*}(u, v^{*}) = F_{K}^{*}(v^{*}) - G_{K}^{*}(v^{*}) + \frac{1}{2\varepsilon} \left\| \frac{\partial G_{K}^{*}(v^{*})}{\partial v^{*}} - u \right\|_{2}^{2} + \frac{1}{2\varepsilon} \left\| \frac{\partial F_{K}^{*}(v^{*})}{\partial v^{*}} - u \right\|_{2}^{2} + \frac{1}{2\varepsilon} \left\| \frac{\partial G_{K}^{*}(v^{*})}{\partial v^{*}} - \frac{\partial F_{K}^{*}(v^{*})}{\partial v^{*}} \right\|_{2}^{2},$$
(20)

we have

$$J(u_0) = \inf_{u \in V} J(u)$$

$$= \inf_{(u,v^*) \in V \times V^*} J_1^*(u,v^*)$$

$$= J_1^*(u_0,v_0^*). \tag{21}$$

Proof. Observe that from the hypotheses and the results and statements of the last subsection

$$J(u_0) = \inf_{u \in V} J(u) = \inf_{v^* \in Y^*} J_K^*(v^*) = J_K^*(v_0^*),$$

where

$$J_K^*(v^*) = F_K^*(v^*) - G_K^*(v^*), \forall v^* \in V^*.$$

Moreover we have

$$J_1^*(u, v^*) \ge J_K^*(v^*), \forall u \in V, \ v^* \in V^*.$$

Also from hypotheses and the last subsection results,

$$u_0 = \frac{\partial F_K^*(v_0^*)}{\partial v^*} = \frac{\partial G_K^*(v_0^*)}{\partial v^*},$$

so that clearly we have

$$J_1^*(u_0, v_0^*) = J_K^*(v_0^*).$$

From these last results, we may infer that

$$J(u_0) = \inf_{u \in V} J(u)$$

$$= \inf_{v^* \in V^*} J_K^*(v^*)$$

$$= J_K^*(v_0^*)$$

$$= \inf_{(u,v^*) \in V \times V^*} J_1^*(u,v^*)$$

$$= J_1^*(u_0,v_0^*). \tag{22}$$

The proof is complete.

Remark 5.2. At this point we highlight that J_1^* has a large region of convexity around the optimal point (u_0, v_0^*) , for K > 0 sufficiently large and corresponding $\varepsilon > 0$ sufficiently small. Indeed, observe that for $v^* \in V^*$,

$$G_K^*(v^*) = \sup_{u \in V} \{ \langle u, v^* \rangle_V - G_K(u) \} = \langle \hat{u}, v^* \rangle_V - G_K(\hat{u})$$

where $\hat{u} \in V$ is such that

$$v^* = \frac{\partial G_K(\hat{u})}{\partial u} = G'(\hat{u}) + K\hat{u}.$$

Taking the variation in v^* in this last equation, we obtain

$$1 = G''(u)\frac{\partial \hat{u}}{\partial v^*} + K\frac{\partial \hat{u}}{\partial v^*},$$

so that

$$\frac{\partial \hat{u}}{\partial v^*} = \frac{1}{G''(u) + K} = \mathcal{O}\left(\frac{1}{K}\right).$$

From this we get

$$\frac{\partial^2 \hat{u}}{\partial (v^*)^2} = -\frac{1}{(G''(u) + K)^2} G'''(u) \frac{\partial \hat{u}}{\partial v^*}$$

$$= -\frac{1}{(G''(u) + K)^3} G'''(u)$$

$$= \mathcal{O}\left(\frac{1}{K^3}\right). \tag{23}$$

On the other hand, from the implicit function theorem

$$\frac{\partial G_K^*(v^*)}{\partial v^*} = u + [v^* - G_K'(\hat{u})] \frac{\partial \hat{u}}{\partial v^*} = u,$$

so that

$$\frac{\partial^2 G_K^*(v^*)}{\partial (v^*)^2} = \frac{\partial \hat{u}}{\partial v^*} = \mathcal{O}\left(\frac{1}{K}\right)$$

and

$$\frac{\partial^3 G_K^*(v^*)}{\partial (v^*)^3} = \frac{\partial^2 \hat{u}}{\partial (v^*)^2} = \mathcal{O}\left(\frac{1}{K^3}\right).$$

Similarly, we may obtain

$$\frac{\partial^2 F_K^*(v^*)}{\partial (v^*)^2} = \mathcal{O}\left(\frac{1}{K}\right)$$

and

$$\frac{\partial^3 F_K^*(v^*)}{\partial (v^*)^3} = \mathcal{O}\left(\frac{1}{K^3}\right).$$

Denoting

$$A = \frac{\partial^2 F_K^*(v_0^*)}{\partial (v^*)^2}$$

and

$$B = \frac{\partial^2 G_K^*(v_0^*)}{\partial (v^*)^2},$$

we have

$$\frac{\partial^2 J_1^*(u_0, v_0^*)}{\partial (v^*)^2} = A - B + \frac{1}{\varepsilon} \left(2A^2 + 2B^2 - 2AB \right),$$
$$\frac{\partial^2 J_1^*(u_0, v_0^*)}{\partial u^2} = \frac{2}{\varepsilon},$$

and

$$\frac{\partial^2 J_1^*(u_0, v_0^*)}{\partial (v^*) \partial u} = -\frac{1}{\varepsilon} (A+B).$$

From this we get

$$\det(\delta^{2} J^{*}(v_{0}^{*}, u_{0})) = \frac{\partial^{2} J_{1}^{*}(u_{0}, v_{0}^{*})}{\partial (v^{*})^{2}} \frac{\partial^{2} J_{1}^{*}(u_{0}, v_{0}^{*})}{\partial u^{2}} - \left[\frac{\partial^{2} J_{1}^{*}(u_{0}, v_{0}^{*})}{\partial (v^{*})\partial u}\right]^{2}$$

$$= 2\frac{A - B}{\varepsilon} + 2\frac{(A - B)^{2}}{\varepsilon^{2}}$$

$$= \mathcal{O}\left(\frac{1}{\varepsilon^{2}}\right)$$

$$\gg \mathbf{0}$$
(24)

about the optimal point (u_0, v_0^*) .

Remark 5.3. Denoting again $Y = Y^* = L^2(\Omega)$, we may also define the functionals $F: V \to \mathbb{R}$ and $G: V \times Y \to \mathbb{R}$, where

$$F(u) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx + \frac{K}{2} \int_{\Omega} u^2 \, dx - \langle u, f \rangle_{L^2}$$

and

$$G(u,v) = -\frac{\alpha}{2} \int_{\Omega} (u^2 - \beta + v)^2 dx + \frac{K}{2} \int_{\Omega} u^2 dx,$$

and the respective polar functionals $F:Y^*\to\mathbb{R}$ and $G^*:[Y^*]^2\to\mathbb{R}$ by

$$F^{*}(v^{*}) = \sup_{u \in V} \{\langle u, v^{*} \rangle_{L^{2}} - F(u)\}$$

$$= \sup_{u \in V} \left\{ \langle u, v^{*} \rangle_{L^{2}} - \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx - \frac{K}{2} \int_{\Omega} u^{2} \, dx + \langle u, f \rangle_{L^{2}} \right\}$$

$$= \frac{1}{2} \int_{\Omega} \frac{(v^{*} + f)^{2}}{-\gamma \nabla^{2} + K} \, dx, \qquad (25)$$

$$G^{*}(v^{*}, v_{0}^{*}) = \sup_{u \in V} \left\{ \inf_{v \in Y} \{ \langle u, v^{*} \rangle_{L^{2}} - G(u, v) \} \right\}$$

$$= \sup_{u \in V} \inf_{v \in L^{2}} \{ \langle u, v^{*} \rangle_{L^{2}} - \langle v, v_{0}^{*} \rangle_{L^{2}}$$

$$+ \frac{\alpha}{2} \int_{\Omega} (u^{2} - \beta + v)^{2} dx - \frac{K}{2} \int_{\Omega} u^{2} dx \right\}$$

$$= -\frac{1}{2} \int_{\Omega} \frac{(v^{*})^{2}}{(2v_{0}^{*} - K)} - \frac{1}{2\alpha} \int_{\Omega} (v_{0}^{*})^{2} dx - \beta \int_{\Omega} v_{0}^{*} dx, \qquad (26)$$

in

$$A^* = \{v_0^* \in Y^* : -2v_0^* + K > K/2, \text{ in } \Omega\}.$$

Finally, we define

$$J_2^*(v^*, v_0^*) = -F^*(v^*) + G^*(v^*, v_0^*).$$

Observe that, at first, this functional is not convex.

However, for an appropriate choice of K > 0, $K_1 > 0$ and $K_2 > 0$, the extended functional $J_3^*: B^* \times A^* \to \mathbb{R}$ may be convex in v^* in a large region about a critical point and concave in v_0^* on $B^* \times A^*$, where

$$B^* = \{ v^* \in Y^* : \|v^*\|_{\infty} \le K_2 \},\$$

and

$$J_{3}^{*}(v^{*}, v_{0}^{*}) = J_{2}^{*}(v^{*}, v_{0}^{*}) + \frac{K_{1}}{2} \int_{\Omega} \left| \frac{\partial J_{2}^{*}(v^{*}, v_{0}^{*})}{\partial v^{*}} \right|^{2} dx$$

$$= -\frac{1}{2} \int_{\Omega} \frac{(v^{*} + f)^{2}}{-\gamma \nabla^{2} + K} dx - \frac{1}{2} \int_{\Omega} \frac{(v^{*})^{2}}{2v_{0}^{*} - K} dx$$

$$-\frac{1}{2\alpha} \int_{\Omega} (v_{0}^{*})^{2} dx - \beta \int_{\Omega} v_{0}^{*} dx$$

$$+ \frac{K_{1}}{2} \int_{\Omega} \left| \frac{v^{*} + f}{-\gamma \nabla^{2} + K} - \frac{v^{*}}{-2v_{0}^{*} + K} \right|^{2} dx.$$
(27)

We highlight the appropriate choice of K, K_1 and $K_2 > 0$ depends on α, β and γ . It is also worth emphasizing the critical points of J_2^* and J_3^* are the same.

6 A convex dual variational formulation for global optimization

In this section, for $\Omega \subset \mathbb{R}^3$ open, bounded, connected and with a regular boundary $\partial\Omega$, we define again $V = W_0^{1,2}(\Omega)$, $Y = Y^* = L^2(\Omega)$ and $F : V \to \mathbb{R}$, $G : V \times Y \to \mathbb{R}$ by

$$F(u) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx + \frac{K}{2} \int_{\Omega} u^2 \, dx - \langle u, f \rangle_{L^2},$$
$$G(u, v) = -\frac{\alpha}{2} \int_{\Omega} (u^2 - \beta + v)^2 \, dx + \frac{K}{2} \int_{\Omega} u^2 \, dx,$$

so that

$$J(u) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx$$

$$+ \frac{\alpha}{2} \int_{\Omega} (u^2 - \beta)^2 \, dx - \langle u, f \rangle_{L^2}$$

$$= F(u) - G(u, 0), \tag{28}$$

where $\gamma > 0$, $\alpha > 0$, $\beta > 0$ and $f \in L^2(\Omega)$.

Define also $F^*: Y^* \to \mathbb{R}$ by

$$F^{*}(v^{*}) = \sup_{u \in V} \{\langle u, v^{*} \rangle_{L^{2}} - F(u) \}$$

$$= \frac{1}{2} \int_{\Omega} \frac{(v^{*} + f)^{2}}{-\gamma \nabla^{2} + K} dx,$$
(29)

$$G^{*}(v^{*}, v_{0}^{*}) = \sup_{u \in V} \left\{ \inf_{v \in Y} \left\{ \langle u, v^{*} \rangle_{L^{2}} - \langle v, v_{0}^{*} \rangle_{L^{2}} - G(u, v) \right\} \right\}$$

$$= -\frac{1}{2} \int_{\Omega} \frac{(v^{*})^{2}}{2v_{0}^{*} - K} dx - \frac{1}{2\alpha} \int_{\Omega} (v_{0}^{*})^{2} dx - \beta \int_{\Omega} v_{0}^{*} dx$$
(30)

if $v_0^* \in B^*$, where

$$B^* = \{ v_0^* \in Y^* : ||v_0^*||_{\infty} < K_1 \},$$

for an appropriate constant $K_1 > 0$.

Moreover define $J^*: [Y^*]^2 \to \mathbb{R}$ by

$$J^*(v^*, v_0^*) = -F^*(v^*) + G^*(v^*, v_0^*),$$

and

$$J_{1}^{*}(v^{*}, v_{0}^{*}) = J^{*}(v^{*}, v_{0}^{*})$$

$$+ \frac{K_{3}}{2} \int_{\Omega} \left(\frac{v^{*} + f}{-\gamma \nabla^{2} + K} - \frac{v^{*}}{-2v_{0}^{*} + K} \right)^{2} dx$$

$$\frac{K_{3}}{2} \int_{\Omega} \left(\frac{v_{0}^{*}}{\alpha} - \left(\frac{v^{*} + f}{-\gamma \nabla^{2} + K} \right)^{2} + \beta \right)^{2} dx.$$

$$J_{3}^{*}(v^{*}) = \sup_{v_{0}^{*} \in B^{*}} J^{*}(v^{*}, v_{0}^{*}),$$

$$(31)$$

Observe that the critical points of J^* and J_3^* are the same. On the other hand, defining

$$D^* = \{ v^* \in Y^* : ||v^*||_{\infty} < K_2, \},$$
$$\tilde{V} = \{ u \in V : ||u||_{\infty} < K_4 \},$$

from the general result in Toland [15], for appropriate K > 0, $K_1 > 0$ and $K_2 > 0$, $K_3 > 0$, $K_4 > 0$, we have

$$\inf_{u \in \tilde{V}} J(u) = \inf_{v^* \in D^*} J_3^*(v^*).$$

Since the critical points of J_1^* correspond in an one to one fashion to critical points of J_3^* , from the Ekeland variational principle we may obtain a minimizing sequence for J_3^* which corresponds to a minimizing sequence for J_1^* so that, we may obtain

$$\inf_{u \in \tilde{V}} J(u) = \inf_{v^* \in D^*} J_3^*(v^*) = \inf_{(v^*, v_0^*) \in D^* \times B^*} J_1^*(v^*, v_0^*).$$

Considering these last definitions and results, we have obtained a proof of the following theorem.

Theorem 6.1. Considering the statements in the last lines in this section, let $\hat{v}^* \in D^*$ be such that

$$J_3^*(\hat{v}^*) = \min_{v^* \in D^*} J_3^*(v^*).$$

Assume $u_0 \in V$ and $\hat{v}_0^* \in Y^*$ such that

$$u_0 = \frac{\hat{v}^* + f}{\gamma \nabla^2 + K}$$

and

$$v_0^* = \alpha(u_0^2 - \beta),$$

are also such that

$$u_0 \in \tilde{V}$$

and

$$v_0^* \in B^*$$
.

Under such hypotheses, we have

$$J_{3}^{*}(\hat{v}^{*}) = \inf_{v^{*} \in D^{*}} J_{3}^{*}(v^{*})$$

$$= \inf_{(v^{*}, v_{0}^{*}) \in D^{*} \times B^{*}} J_{1}^{*}(v^{*}, v_{0}^{*})$$

$$= J_{1}^{*}(\hat{v}^{*}, \hat{v}_{0}^{*})$$

$$= J(u_{0})$$

$$= \inf_{u \in \tilde{V}} J(u).$$
(32)

Remark 6.2. We highlight such a duality principle concerns a convex dual variational formulation suitable for a global optimization of the primal formulation, in the specific sense that, it has a large region of convexity around a critical point. Finally, we highlight such a dual formulation is applicable to a large class of models in physics and engineering.

7 A final convex dual variational formulation

In this section, again for $\Omega \subset \mathbb{R}^3$ an open, bounded, connected set with a regular (Lipschitzian) boundary $\partial\Omega$, $\gamma>0$, $\alpha>0$, $\beta>0$ and $f\in L^2(\Omega)$, we denote $F_1:V\times Y\to\mathbb{R}$, $F_2:V\to\mathbb{R}$ and $G:V\times Y\to\mathbb{R}$ by

$$F_{1}(u, v_{0}^{*}) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx - \frac{K}{2} \int_{\Omega} u^{2} \, dx + \frac{K_{1}}{2} \int_{\Omega} (-\gamma \nabla^{2} u + 2v_{0}^{*} u - f)^{2} \, dx + \frac{K_{2}}{2} \int_{\Omega} u^{2} \, dx,$$
(33)

$$F_2(u) = \frac{K_2}{2} \int_{\Omega} u^2 dx + \langle u, f \rangle_{L^2},$$

and

$$G(u,v) = \frac{\alpha}{2} \int_{\Omega} (u^2 - \beta + v)^2 dx + \frac{K}{2} \int_{\Omega} u^2 dx.$$

We define also

$$J_1(u, v_0^*) = F_1(u, v_0^*) - F_2(u) + G(u, 0),$$

$$J(u) = \frac{\gamma}{2} \int_{\Omega} \nabla u \cdot \nabla u \, dx + \frac{\alpha}{2} \int_{\Omega} (u^2 - \beta)^2 \, dx - \langle u, f \rangle_{L^2},$$

and $F_1^*:[Y^*]^3\to\mathbb{R},\,F_2^*:Y^*\to\mathbb{R},\,\text{and}\,\,G^*:[Y^*]^2\to\mathbb{R},\,\text{by}$

$$F_1^*(v_2^*, v_1^*, v_0^*) = \sup_{u \in V} \{ \langle u, v_1^* + v_2^* \rangle_{L^2} - F_1(u, v_0^*) \}$$

$$= \frac{1}{2} \int_{\Omega} \frac{\left(v_1^* + v_2^* + K_1(-\gamma \nabla^2 + 2v_0^*)f \right)^2}{(-\gamma \nabla^2 - K + K_2 + K_1(-\gamma \nabla^2 + 2v_0^*)^2)} dx$$

$$-\frac{K_1}{2} \int_{\Omega} f^2 dx, \tag{34}$$

$$F_2^*(v_2^*) = \sup_{u \in V} \{\langle u, v_2^* \rangle_{L^2} - F_2(u)\}$$

$$= \frac{1}{2K_2} \int_{\Omega} (v_2^*)^2 dx, \qquad (35)$$

and

$$G^{*}(v_{1}^{*}, v_{0}^{*}) = \sup_{(u,v) \in V \times Y} \{\langle u, v_{1}^{*} \rangle_{L^{2}} - \langle v, v_{0}^{*} \rangle_{L^{2}} - G(u, v)\}$$

$$= \frac{1}{2} \int_{\Omega} \frac{(v_{1}^{*})^{2}}{2v_{0}^{*} + K} dx + \frac{1}{2\alpha} \int_{\Omega} (v_{0}^{*})^{2} dx$$

$$+\beta \int_{\Omega} v_{0}^{*} dx$$
(36)

if $v_0^* \in B^*$ where

$$B^* = \{ v_0^* \in Y^* : ||v_0^*||_{\infty} \le K/2 \}.$$

Finally, we also define $J_1^*: [Y^*]^2 \times B^* \to \mathbb{R}$

$$J_1^*(v_2^*, v_1^*, v_0^*) = -F_1^*(v_2^*, v_1^*, v_0^*) + F_2^*(v_2^*) - G^*(v_1^*, v_0^*).$$

By computing $\delta^2 J_1^*(v_2^*, v_1^*, v_0^*)$ we may obtain that for appropriate K > 0, $K_1 > 0$, $K_2 > 0$, J_1^* in convex in v_2^* on Y^* and it is concave in (v_1^*, v_0^*) around any critical point.

Considering such statements and definitions, we may prove the following theorem.

Theorem 7.1. Let $(\hat{v}_{2}^{*}, \hat{v}_{1}^{*}, \hat{v}_{0}^{*}) \in Y^{*} \times Y^{*} \times B^{*}$ be such that

$$\delta J_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*) = \mathbf{0}$$

and $u_0 \in V$ be such that

$$u_0 = \frac{\hat{v}_1^* + \hat{v}_2^* + K_1(-\gamma \nabla^2 + 2v_0^*)f}{K_2 - K - \gamma \nabla^2 + K_1(-\gamma \nabla^2 + 2\hat{v}_0^*)^2}.$$

Under such hypotheses, we have

$$\delta J(u_0) = \mathbf{0},$$

so that

$$J(u_0) = \inf_{u \in V} \left\{ J(u) + \frac{K_1}{2} \int_{\Omega} (-\gamma \nabla^2 u + 2\hat{v}_0^* u - f)^2 dx \right\}$$

$$= \inf_{v_2^* \in Y^*} \left\{ \sup_{(v_1^*, v_0^*) \in Y^* \times B^*} J_1^*(v_2^*, v_1^*, v_0^*) \right\}$$

$$= J_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*). \tag{37}$$

Proof. Observe that $\delta J_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*) = \mathbf{0}$ so that, since J_1^* is convex in v_2^* on Y^* and concave in (v_1^*, v_0^*) on $Y^* \times B^*$, we obtain

$$J_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*) = \inf_{v_2^* \in Y^*} \left\{ \sup_{(v_1^*, v_0^*) \in Y^* \times B^*} J_1^*(v_2^*, v_1^*, v_0^*) \right\}.$$

Now we are going to show that

$$\delta J(u_0) = \mathbf{0}.$$

From

$$\frac{\partial J_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*)}{\partial v_2^*} = \mathbf{0},$$

we have

$$-u_0 + \frac{\hat{v}_2^*}{K_2} = 0,$$

and thus

$$\hat{v}_2^* = K_2 u_0.$$

From

$$\frac{\partial J_1^*(\hat{v}_2^*,\hat{v}_1^*,\hat{v}_0^*)}{\partial v_1^*} = \mathbf{0},$$

we obtain

$$-u_0 - \frac{\hat{v}_1^* - f}{2\hat{v}_0^* + K} = 0,$$

and thus

$$\hat{v}_1^* = -2\hat{v}_0^* u_0 - Ku_0 + f.$$

Finally, denoting

$$D = -\gamma \nabla^2 u_0 + 2\hat{v}_0^* u_0 - f,$$

from

$$\frac{\partial J_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*)}{\partial v_0^*} = \mathbf{0},$$

we have

$$-2Du_0 + u_0^2 - \frac{\hat{v}_0^*}{\alpha} - \beta = 0,$$

so that

$$\hat{v}_0^* = \alpha (u_0^2 - \beta - 2Du_0). \tag{38}$$

Observe now that

$$\hat{v}_1^* + \hat{v}_2^* + K_1(-\gamma \nabla^2 + 2\hat{v}_0^*)f = (K_2 - K - \gamma \nabla^2 + K_1(-\gamma \nabla^2 + 2\hat{v}_0^*)^2)u_0$$

so that

$$K_2 u_0 - 2\hat{v}_0 u_0 - K u_0 + f$$

$$= K_2 u_0 - K u_0 - \gamma \nabla^2 u_0 + K_1 (-\gamma \nabla^2 + 2\hat{v}_0^*) (-\gamma \nabla^2 u_0 + 2\hat{v}_0^* u_0 - f).$$
(39)

The solution for this last system of equations (38) and (39) is obtained through the relations

$$\hat{v}_0^* = \alpha(u_0^2 - \beta)$$

and

$$-\gamma \nabla^2 u_0 + 2\hat{v}_0^* u_0 - f = D = 0,$$

so that

$$\delta J(u_0) = -\gamma \nabla^2 u_0 + 2\alpha (u_0^2 - \beta) u_0 - f = 0$$

and

$$\delta \left\{ J(u_0) + \frac{K_1}{2} \int_{\Omega} (-\gamma \nabla^2 u_0 + 2\hat{v}_0^* u_0 - f)^2 dx \right\} = 0,$$

and hence, from the concerning convexity in u on V,

$$J(u_0) = \min_{u \in V} \left\{ J(u) + \frac{K_1}{2} \int_{\Omega} (-\gamma \nabla^2 u + 2\hat{v}_0^* u - f)^2 dx \right\}.$$

Moreover, from the Legendre transform properties

$$F_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*) = \langle u_0, \hat{v}_2^* + \hat{v}_1^* \rangle_{L^2} - F_1(u_0, \hat{v}_0^*),$$

$$F_2^*(\hat{v}_2^*) = \langle u_0, \hat{v}_2^* \rangle_{L^2} - F_2(u_0),$$

$$G^*(\hat{v}_1^*, \hat{v}_0^*) = -\langle u_0, \hat{v}_1^* \rangle_{L^2} - \langle 0, \hat{v}_0^* \rangle_{L^2} - G(u_0, 0),$$

so that

$$J_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*) = -F_1^*(\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*) + F_2^*(\hat{v}_2^*) - G^*(\hat{v}_1^*, \hat{v}_0^*)$$

$$= F_1(u_0, \hat{v}_0^*) - F_2(u_0) + G(u_0, 0)$$

$$= J(u_0). \tag{40}$$

Joining the pieces, we have got

$$J(u_0) = \inf_{u \in V} \left\{ J(u) + \frac{K_1}{2} \int_{\Omega} (-\gamma \nabla^2 u + 2\hat{v}_0^* u - f)^2 dx \right\}$$

$$= \inf_{v_2^* \in Y^*} \left\{ \sup_{(v_1^*, v_0^*) \in Y^* \times B^*} J_1^* (v_2^*, v_1^*, v_0^*) \right\}$$

$$= J_1^* (\hat{v}_2^*, \hat{v}_1^*, \hat{v}_0^*). \tag{41}$$

The proof is complete.

8 A related numerical computation through the generalized method of lines

We start by recalling that the generalized method of lines was originally introduced in the book entitled "Topics on Functional Analysis, Calculus of Variations and Duality" [7], published in 2011.

Indeed, the present results are extensions and applications of previous ones which have been published since 2011, in books and articles such as [7, 8, 9, 5]. About the Sobolev spaces involved we would mention [1]. Concerning the applications, related models in physics are addressed in [4, 11].

We also emphasize that, in such a method, the domain of the partial differential equation in question is discretized in lines (or more generally, in curves) and the concerning solution is written on these lines as functions of boundary conditions and the domain boundary shape.

In fact, in its previous format, this method consists of an application of a kind of a partial finite differences procedure combined with the Banach fixed point theorem to obtain the relation between two adjacent lines (or curves).

In the present article, we propose an improvement concerning the way we truncate the series solution obtained through an application of the Banach fixed point theorem to find the relation between two adjacent lines. The results obtained are very good even as a typical parameter $\varepsilon > 0$ is very small.

In the next lines and sections we develop in details such a numerical procedure.

8.1 About a concerning improvement for the generalized method of lines

Let $\Omega \subset \mathbb{R}^2$ where

$$\Omega = \{ (r, \theta) \in \mathbb{R}^2 : 1 \le r \le 2, \ 0 \le \theta \le 2\pi \}.$$

Consider the problem of solving the partial differential equation

$$\begin{cases}
-\varepsilon \left(\frac{\partial^2 u}{\partial r^2} + \frac{1}{r} \frac{\partial u}{\partial r} + \frac{1}{r^2} \frac{\partial^2 u}{\partial \theta^2} \right) + \alpha u^3 - \beta u = f, & \text{in } \Omega, \\
u = u_0(\theta), & \text{on } \partial \Omega_1, \\
u = u_f(\theta), & \text{on } \partial \Omega_2.
\end{cases} \tag{42}$$

Here

$$\Omega = \{ (r, \theta) \in \mathbb{R}^2 : 1 \le r \le 2, \ 0 \le \theta \le 2\pi \},$$
$$\partial \Omega_1 = \{ (1, \theta) \in \mathbb{R}^2 : 0 \le \theta \le 2\pi \},$$
$$\partial \Omega_2 = \{ (2, \theta) \in \mathbb{R}^2 : 0 \le \theta \le 2\pi \},$$

 $\varepsilon > 0$, $\alpha > 0$, $\beta > 0$, and $f \equiv 1$, on Ω .

In a partial finite differences scheme, such a system stands for

$$-\varepsilon \left(\frac{u_{n+1} - 2u_n + u_{n-1}}{d^2} + \frac{1}{t_n} \frac{u_n - u_{n-1}}{d} + \frac{1}{t_n^2} \frac{\partial^2 u_n}{\partial \theta^2} \right) + \alpha u_n^3 - \beta u_n = f_n,$$

 $\forall n \in \{1, \dots, N-1\}$, with the boundary conditions

$$u_0 = 0,$$

and

$$u_N = 0.$$

Here N is the number of lines and d = 1/N.

In particular, for n = 1 we have

$$-\varepsilon \left(\frac{u_2 - 2u_1 + u_0}{d^2} + \frac{1}{t_1} \frac{(u_1 - u_0)}{d} + \frac{1}{t_1^2} \frac{\partial^2 u_1}{\partial \theta^2} \right) + \alpha u_1^3 - \beta u_1 = f_1,$$

so that

$$u_1 = \left(u_2 + u_1 + u_0 + \frac{1}{t_1}(u_1 - u_0) d + \frac{1}{t_1^2} \frac{\partial^2 u_1}{\partial \theta^2} d^2 + (-\alpha u_1^3 + \beta u_1 - f_1) \frac{d^2}{\varepsilon}\right) / 3.0,$$

We solve this last equation through the Banach fixed point theorem, obtaining u_1 as a function of u_2 .

Indeed, we may set

$$u_1^0 = u_2$$

and

$$u_1^{k+1} = \left(u_2 + u_1^k + u_0 + \frac{1}{t_1}(u_1^k - u_0) d + \frac{1}{t_1^2} \frac{\partial^2 u_1^k}{\partial \theta^2} d^2 + (-\alpha(u_1^k)^3 + \beta u_1^k - f_1) \frac{d^2}{\varepsilon}\right) / 3.0, \tag{43}$$

 $\forall k \in \mathbb{N}.$

Thus, we may obtain

$$u_1 = \lim_{k \to \infty} u_1^k \equiv H_1(u_2, u_0).$$

Similarly, for n=2, we have

$$u_{2} = \left(u_{3} + u_{2} + H_{1}(u_{2}, u_{0}) + \frac{1}{t_{1}}(u_{2} - H_{1}(u_{2}, u_{0})) d + \frac{1}{t_{1}^{2}} \frac{\partial^{2} u_{2}}{\partial \theta^{2}} d^{2} + (-\alpha u_{2}^{3} + \beta u_{2} - f_{2}) \frac{d^{2}}{\varepsilon}\right) / 3.0,$$

$$(44)$$

We solve this last equation through the Banach fixed point theorem, obtaining u_2 as a function of u_3 and u_0 .

Indeed, we may set

$$u_2^0 = u_3$$

and

$$u_2^{k+1} = \left(u_3 + u_2^k + H_1(u_2^k, u_0) + \frac{1}{t_2}(u_2^k - H_1(u_2^k, u_0)) d + \frac{1}{t_2^2} \frac{\partial^2 u_2^k}{\partial \theta^2} d^2 + (-\alpha(u_2^k)^3 + \beta u_2^k - f_2) \frac{d^2}{\varepsilon}\right) / 3.0, \tag{45}$$

 $\forall k \in \mathbb{N}.$

Thus, we may obtain

$$u_2 = \lim_{k \to \infty} u_2^k \equiv H_2(u_3, u_0).$$

Now reasoning inductively, having

$$u_{n-1} = H_{n-1}(u_n, u_0),$$

we may get

$$u_{n} = \left(u_{n+1} + u_{n} + H_{n-1}(u_{n}, u_{0}) + \frac{1}{t_{n}}(u_{n} - H_{n-1}(u_{n}, u_{0})) d + \frac{1}{t_{n}^{2}} \frac{\partial^{2} u_{n}}{\partial \theta^{2}} d^{2} + (-\alpha u_{n}^{3} + \beta u_{n} - f_{n}) \frac{d^{2}}{\varepsilon}\right) / 3.0,$$

$$(46)$$

We solve this last equation through the Banach fixed point theorem, obtaining u_n as a function of u_{n+1} and u_0 .

Indeed, we may set

$$u_n^0 = u_{n+1}$$

and

$$u_n^{k+1} = \left(u_{n+1} + u_n^k + H_{n-1}(u_n^k, u_0) + \frac{1}{t_n}(u_n^k - H_{n-1}(u_n^k, u_0)) d + \frac{1}{t_n^2} \frac{\partial^2 u_n^k}{\partial \theta^2} d^2 + (-\alpha(u_n^k)^3 + \beta u_n^k - f_n) \frac{d^2}{\varepsilon}\right) / 3.0,$$

$$(47)$$

 $\forall k \in \mathbb{N}.$

Thus, we may obtain

$$u_n = \lim_{k \to \infty} u_n^k \equiv H_n(u_{n+1}, u_0).$$

We have obtained $u_n = H_n(u_{n+1}, u_0), \forall n \in \{1, \dots, N-1\}$. In particular, $u_N = u_f(\theta)$, so that we may obtain

$$u_{N-1} = H_{N-1}(u_N, u_0) = H_{N-1}(0) \equiv F_{N-1}(u_N, u_0) = F_{N-1}(u_f(\theta), u_0(\theta)).$$

Similarly,

$$u_{N-2} = H_{N-2}(u_{N-1}, u_0) = H_{N-2}(H_{N-1}(u_N, u_0)) = F_{N-2}(u_N, u_0) = F_{N-1}(u_f(\theta), u_0(\theta)),$$

an so on, up to obtaining

$$u_1 = H_1(u_2) \equiv F_1(u_N, u_0) = F_1(u_f(\theta), u_0(\theta)).$$

The problem is then approximately solved.

8.2 Software in Mathematica for solving such an equation

We recall that the equation to be solved is a Ginzburg-Landau type one, where

$$\begin{cases}
-\varepsilon \left(\frac{\partial^2 u}{\partial r^2} + \frac{1}{r} \frac{\partial u}{\partial r} + \frac{1}{r^2} \frac{\partial^2 u}{\partial \theta^2} \right) + \alpha u^3 - \beta u = f, & \text{in } \Omega, \\
u = 0, & \text{on } \partial \Omega_1, \\
u = u_f(\theta), & \text{on } \partial \Omega_2.
\end{cases} \tag{48}$$

Here

$$\Omega = \{ (r, \theta) \in \mathbb{R}^2 : 1 \le r \le 2, \ 0 \le \theta \le 2\pi \},
\partial \Omega_1 = \{ (1, \theta) \in \mathbb{R}^2 : 0 \le \theta \le 2\pi \},
\partial \Omega_2 = \{ (2, \theta) \in \mathbb{R}^2 : 0 \le \theta \le 2\pi \},$$

 $\varepsilon > 0, \ \alpha > 0, \beta > 0,$ and $f \equiv 1,$ on Ω . In a partial finite differences scheme, such a system stands for

$$-\varepsilon \left(\frac{u_{n+1} - 2u_n + u_{n-1}}{d^2} + \frac{1}{t_n} \frac{u_n - u_{n-1}}{d} + \frac{1}{t_n^2} \frac{\partial^2 u_n}{\partial \theta^2} \right) + \alpha u_n^3 - \beta u_n = f_n,$$

 $\forall n \in \{1, \dots, N-1\}$, with the boundary conditions

$$u_0 = 0,$$

and

$$u_N = u_f[x].$$

Here N is the number of lines and d = 1/N.

At this point we present the concerning software for an approximate solution.

Such a software is for N = 10 (10 lines) and $u_0[x] = 0$..

- 1. $m_8 = 10$; $(N = 10 \ lines)$
- 2. d = 1/m8;
- 3. $e_1 = 0.1$; $(\varepsilon = 0.1)$
- 4. A = 1.0;
- 5. B = 1.0:
- 6. $For[i = 1, i < m8, i + +, f[i] = 1.0]; (f \equiv 1, \text{ on } \Omega)$
- 7. a = 0.0;
- 8. For[i = 1, i < m8, i + +, Clear[b, u]; t[i] = 1 + i * d; $b[x_{-}] = u[i + 1][x];$

```
9. For[k = 1, k < 30, k + +, \text{ (we have fixed the number of iterations)}]
     z = \left(u[i+1][x] + b[x] + a + \tfrac{1}{t[i]}(b[x]-a)*d\right.
     \left. + \frac{1}{t[i]^2} D[b[x], \{x,2\}] * d^2 + (-A*b[x]^3 + B*u[x] + f[i]) * \frac{d^2}{e_1} \right) / 3.0;
     Series[z, \{u[i+1][x], 0, 3\}, \{u[i+1]'[x], 0, 1\}, \{u[i+1]''[x], 0, 1\},
     {u[i+1]'''[x], 0, 0}, {u[i+1]''''[x], 0, 0};
     z = Normal[z],
     z = Expand[z];
     b[x_{-}] = z];
10. a_1[i] = z;
11. Clear[b];
12. u[i+1][x_{-}] = b[x];
13. a = a_1[i];
14. b[x_{-}] = u_{f}[x];
15. For[i = 1, i < m8, i + +,
     A_1 = a_1[m8 - i];
     A_1 = Series[A_1, \{u_f[x], 0, 3\}, \{u_f'[x], 0, 1\}, \{u_f''[x], 0, 1\}, \{u_f'''[x], 0, 0\}, \{u_f''''[x], 0, 0\}];
     A_1 = Normal[A_1];
     A_1 = Expand[A_1];
     u[m8 - i][x_{-}] = A_1;
     b[x_{-}] = A_{1};
     Print[u[m8/2][x]];
   ***********
```

The numerical expressions for the solutions of the concerning N=10 lines are given by

$$u[1][x] = 0.47352 + 0.00691u_f[x] - 0.00459u_f[x]^2 + 0.00265u_f[x]^3 + 0.00039(u_f'')[x] - 0.00058u_f[x](u_f'')[x] + 0.00050u_f[x]^2(u_f'')[x] - 0.000181213u_f[x]^3(u_f'')[x]$$
(49)

$$u[2][x] = 0.76763 + 0.01301u_f[x] - 0.00863u_f[x]^2 + 0.00497u_f[x]^3 + 0.00068(u_f'')[x] - 0.00103u_f[x](u_f'')[x] + 0.00088u_f[x]^2(u_f'')[x] - 0.00034u_f[x]^3(u_f'')[x]$$
(50)

$$u[3][x] = 0.91329 + 0.02034u_f[x] - 0.01342u_f[x]^2 + 0.00768u_f[x]^3 + 0.00095(u_f'')[x] - 0.00144u_f[x](u_f'')[x] + 0.00122u_f[x]^2(u_f'')[x] - 0.00051u_f[x]^3(u_f'')[x]$$
(51)

$$u[4][x] = 0.97125 + 0.03623u_f[x] - 0.02328u_f[x]^2 + 0.01289u_f[x]^3 + 0.00147331(u_f'')[x] - 0.00223u_f[x](u_f'')[x] + 0.00182u_f[x]^2(u_f'')[x] - 0.00074u_f[x]^3(u_f'')[x]$$
(52)

$$u[5][x] = 1.01736 + 0.09242u_f[x] - 0.05110u_f[x]^2 + 0.02387u_f[x]^3 + 0.00211(u_f'')[x] -0.00378u_f[x](u_f'')[x] + 0.00292u_f[x]^2(u_f'')[x] - 0.00132u_f[x]^3(u_f'')[x]$$
(53)

$$u[6][x] = 1.02549 + 0.21039u_f[x] - 0.09374u_f[x]^2 + 0.03422u_f[x]^3 + 0.00147(u_f'')[x] - 0.00634u_f[x](u_f'')[x] + 0.00467u_f[x]^2(u_f'')[x] - 0.00200u_f[x]^3(u_f'')[x]$$
(54)

$$u[7][x] = 0.93854 + 0.36459u_f[x] - 0.14232u_f[x]^2 + 0.04058u_f[x]^3 + 0.00259(u_f'')[x] -0.00747373u_f[x](u_f'')[x] + 0.0047969u_f[x]^2(u_f'')[x] - 0.00194u_f[x]^3(u_f'')[x] (55)$$

$$u[8][x] = 0.74649 + 0.57201u_f[x] - 0.17293u_f[x]^2 + 0.02791u_f[x]^3 + 0.00353(u_f'')[x] - 0.00658u_f[x](u_f'')[x] + 0.00407u_f[x]^2(u_f'')[x] - 0.00172u_f[x]^3(u_f'')[x]$$
(56)

$$u[9][x] = 0.43257 + 0.81004u_f[x] - 0.13080u_f[x]^2 + 0.00042u_f[x]^3 + 0.00294(u_f'')[x] - 0.00398u_f[x](u_f'')[x] + 0.00222u_f[x]^2(u_f'')[x] - 0.00066u_f[x]^3(u_f'')[x]$$
(57)

9 Conclusion

In the first part of this article we develop duality principles for non-convex variational optimization. In the final concerning sections we propose dual convex formulations suitable for a large class of models in physics and engineering. In the last article section, we present an advance concerning the computation of a solution for a partial differential equation through the generalized method of lines. In particular, in its previous versions, we used to truncate the series in d^2 however, we have realized the results are much better by taking line solutions in series for $u_f[x]$ and its derivatives, as it is indicated in the present software.

This is a little difference concerning the previous procedure, but with a great result improvement as the parameter $\varepsilon > 0$ is small.

Indeed, with a sufficiently large N (number of lines), we may obtain very good results even as $\varepsilon > 0$ is very small.

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