

Seminar on Numerical Analysis

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A General Superlinear Convergence Result

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Introduction

Motivation

A General Superlinear Convergence Result

Next Talk

We will look at non-linear operator equation

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1. Equation (1) can also be nonsmooth.
2. We will look at the minimum requirements for the operator G .

Application to Obstacle Problem

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Problem Description: Find the vertical displacement $u(\cdot)$ of a membrane Ω which is loaded by vertical forces $f(\cdot)$ (force density), fixed on the boundary $\Gamma = \delta\Omega$ and located above an obstacle described by $g(\cdot)$

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Constrained Minimization Problem: Find

$$u \in U : \mathfrak{J}(u) = \inf_{v \in U} \mathfrak{J}(v), \quad (2)$$

where, $U := \{v \in H_0^1(\Omega) : v(x) \geq g(x), \forall x \in \Omega\}$ with,
 $\mathfrak{J}(v) := \frac{1}{2} \int_{\Omega} |\nabla v|^2 dx - \int_{\Omega} f v dx$ and given $f \in L^2(\Omega)$,
 $g \in \{v \in H^1(\Omega) : v(x) \leq 0, \forall x \in \Gamma\}$

The minimization problem is equivalent to the Variational Inequality(VI): Find $u \in U$:

$$\int_{\Omega} \nabla^T u \nabla (v - u) dx \geq \int_{\Omega} f(v - u) dx, \quad \forall v \in U \quad (3)$$

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This Variational Inequality is equivalent to the nonlinear fixed point equation. Find $u \in X$:

$$u = B_g(u) := P[(I - \rho \mathcal{J}^{-1} A)u + \rho \mathcal{J}^{-1} f] \quad (4)$$

in X

Application to Optimal Control

Consider the following optimal control problem:

$$\min_{y \in H_0^1(\Omega), u \in L^2(\Omega)} \mathfrak{J}(y, u) := \frac{1}{2} \|y - y_d\|_{L^2(\Omega)}^2 + \frac{\alpha}{2} \|u\|_{L^2(\Omega)}^2 \quad (5)$$

such that $Ay = u$, $\beta_l \leq u \leq \beta_r$

Here, $y \in H_0^1(\Omega)$ is the state, which is defined in Ω

$\Omega \subset \mathbb{R}^n$ is open bounded domain,

$u \in L^2(\Omega)$ is the control,

$A : H_0^1(\Omega) \rightarrow H^{-1}(\Omega) = H_0^1(\Omega)^*$ is a linear elliptic partial differential operator

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This can be rewritten as

$$\Phi(u) = u - P_{[\beta_l, \beta_r]}(u - \theta \nabla \hat{\mathfrak{J}}(u)) = 0. \quad (7)$$

Algorithm: Generalized Newton's Method

Consider the operator equation (1) with $G : X \rightarrow Y$, X, Y Banach spaces. Algorithm for Generalized Newton's Method:

1. Choose $x^0 \in X$
For $k = 0, 1, 2, \dots$
2. Choose an invertible operator $M_k \in L(X, Y)$.
3. Obtain s^k by solving

$$M_k s^k = -G(x^k)$$

and set $x^{k+1} = x^k + s^k$.

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1. (x^k) converges q -linearly to \bar{x} with rate $\gamma \in (0, 1)$ iff

$$\|M_k^{-1}(G(\bar{x} + d^k) - G(\bar{x}) - M_k d^k)\|_X \leq \gamma \|d^k\|_X \quad (8)$$

$\forall k$ with $\|d^k\|_X$ sufficiently small.

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2. (x^k) converges q -superlinearly to \bar{x} iff

$$\|M_k^{-1}(G(\bar{x} + d^k) - G(\bar{x}) - M_k d^k)\|_X = o(\|d^k\|_X) \quad (9)$$

for $\|d^k\|_X \rightarrow 0$

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3. (x^k) converges q -order $1 + \alpha > 1$ to \bar{x} iff

$$\|M_k^{-1}(G(\bar{x} + d^k) - G(\bar{x}) - M_k d^k)\|_X = O(\|d^k\|_X^{1+\alpha}) \quad (10)$$

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1. Regularity condition:

$$\|M_k^{-1}\|_{Y \rightarrow X} \leq C, \quad \forall k \geq 0. \quad (11)$$

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1. Regularity condition:

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2. Approximation condition:

$$\|G(\bar{x} + d^k) - G(\bar{x}) - M_k d^k\|_Y = o(\|d^k\|_X)$$

$$\text{or } \|G(\bar{x} + d^k) - G(\bar{x}) - M_k d^k\|_Y = O(\|d^k\|_X^{1+\alpha})$$

Theorem 1. Consider the operator equation (1) with $G : X \rightarrow Y$, where X and Y are Banach spaces. Let (x^k) be generated by the generalized Newton method. Then:

1. If x^0 is sufficiently close to \bar{x} and (8) holds then $x^k \rightarrow \bar{x}$ q -linearly with rate γ .
2. If x^0 is sufficiently close to \bar{x} and (9) holds then $x^k \rightarrow \bar{x}$ q -superlinearly.
3. If x^0 is sufficiently close to \bar{x} and (10) holds then $x^k \rightarrow \bar{x}$ q -superlinearly with order $1 + \alpha$.

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2. The fast convergence of the generalized Newton's method is not affected if the system is solved inexactly and the accuracy of the solution is controlled suitably.

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3. The regularity condition then reads

$$\|G'(x^k)^{-1}\|_{Y \rightarrow X} \leq C, \quad \forall k \geq 0$$

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2. If G' is α -order Hölder continuous near \bar{x} , we can prove approximation condition of order $1 + \alpha$.

Finally we have the following

Corollary 2 Let $G : X \rightarrow Y$ be a continuously F -differential operator between Banach spaces and assume that $G'(\bar{x})$ is continuously invertible at the solution \bar{x} . Then Newton's method (i.e. Algorithm for Generalized Newton's Method with $M_k = G'(x^k)$ for all k) converges locally q -superlinearly. If, in addition, G' is α -order Hölder continuous near \bar{x} , the order of convergence is $1 + \alpha$.

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2. Definition of Semismoothness.
3. Algorithm for Semismooth Newton Method.
4. Convergence Results.

THANK YOU!!!