

A Newton-type Method for Non-smooth Multi-objective Optimization

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Main Results

Given $F : \mathbb{R}^m \rightarrow \mathbb{R}^n$, $F(x) = [f_1(x), \dots, f_n(x)]$, with **Newton differentiable Jacobian** (e.g. piecewise \mathcal{C}^2),

Problem

Find

$$\operatorname{argmin}_{x \in \mathbb{R}^m}^* F(x) \quad (\text{MOP})$$

Result

The **Newton-type algorithm** converges **superlinearly**

Outline

1. What is argmin^* ?
2. Under-determined Systems of Equations
3. Non-smooth Equation
4. Constraints
5. Main Result

Efficient Points

No total order on \mathbb{R}^m

1. (local) **strict Pareto minimizer**:

$$\forall y(\in U) : F(y) \leq F(\bar{x}) \implies F(y) = F(\bar{x})$$

2. (local) **weak Pareto minimizer**: $\nexists y(\in U) : F(y) < F(\bar{x})$

3. (local) **weakly efficient** $\exists \sigma \in \Delta^{n^1}$, \bar{x} is a (local) minimizer of $\sigma^T F(x)$

Aim to replace (MOP) with an equation

Let $\sigma \in \Delta^n$, $G(x) = \sigma^T \nabla F(x)$ (**first order scalarization**)

NOT good: $F(x) = [x_1^2 - x_2, x_2]$ only **one** good guess for σ

¹ $\Delta^n = \{x \in [0, 1]^n \mid \mathbf{1}^T x = 1\}$

First Order Equation

Given $F : \mathbb{R}^m \rightarrow \mathbb{R}^n$, $F(x) = [f_1(x), \dots, f_n(x)]$, define

$$G : \mathbb{R}^m \times \mathbb{R}^n \rightarrow \mathbb{R}^n, \quad G(x, \sigma) = \sigma^T \nabla F(x)$$

Problem

Find, $\bar{x} \in \mathbb{R}^m, \bar{\sigma} \in \Delta^n$

$$G(\bar{x}, \bar{\sigma}) = 0 \quad (\text{FOE})$$

Under-determined Equations

Given $G : \mathbb{R}^m \rightarrow \mathbb{R}^n$ in \mathcal{C}^1 with $m > n$ Find, $\bar{x} \in \mathbb{R}^m$

$$G(\bar{x}) = 0 \quad (\text{UDE})$$

Newton's method: linearize (UDE) and solve

$$\mathcal{A}(x) = \{y \in \mathbb{R}^m \mid G(x) + \nabla G(x)^T (y - x) = 0\} \quad (\text{EUDE})$$

$\mathcal{A}(x)$ is **not** a singleton

Under-determined Equations

We can project on $\mathcal{A}(x)$

$$x^{k+1} = P_{\mathcal{A}(x^k)} x^k$$

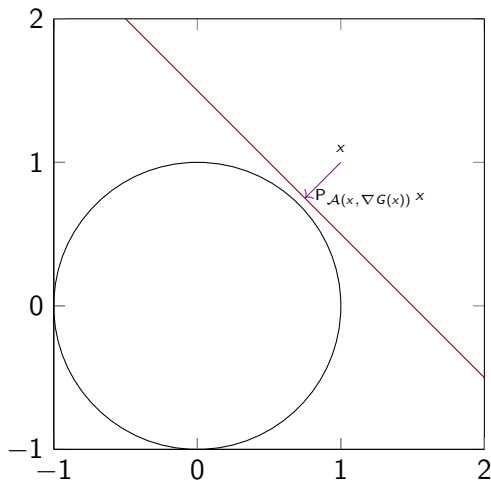
$$x^{k+1} = \underset{G(x) + \nabla G(x^k)^T (x - x^k) = 0}{\operatorname{argmin}} \|x - x^k\|^2$$

First order optimality ($\nabla G(x^k)$ full rank)

$$x^{k+1} = x^k - \nabla G(x^k)^+ \nabla G(x^k)$$

$$^2H^+ = (HH^T)^{-1}H$$

Under-determined Equations



Newton Differentiability

Definition

$G : \mathbb{R}^m \rightarrow \mathbb{R}^n$ is *pointwise Newton differentiable at \bar{x}* if there is $\mathcal{H}G : \mathbb{R}^m \rightrightarrows \mathbb{R}^{n \times m}$ with

$$\lim_{x \rightarrow \bar{x}} \sup_{H \in \mathcal{H}G(x)} \frac{\|G(x) - G(\bar{x}) - H(x - \bar{x})\|}{\|x - \bar{x}\|} = 0$$

Definition

$G : \mathbb{R}^m \rightarrow \mathbb{R}^n$ is *uniformly Newton differentiable on V* if there is $\mathcal{H}G : \mathbb{R}^m \rightrightarrows \mathbb{R}^{n \times m}$ such that $\forall \varepsilon > 0 \exists \delta, \forall x \in \mathbb{R}^m, \forall y \in V$ with $\|x - y\| \leq \delta$,

$$\sup_{H \in \mathcal{H}G(x)} \frac{\|G(x) - G(y) - H(x - y)\|}{\|x - y\|} < \varepsilon.$$

Examples

Proposition

$G : \mathbb{R}^m \rightarrow \mathbb{R}^n$, K compact, $G \in \mathcal{C}^1(K)$, then G is uniformly Newton differentiable on K with a Newton differential

$$\mathcal{H}(x) := \{\nabla G(x)^T\}$$

Definition

$G : \mathbb{R}^m \rightarrow \mathbb{R}^n$ is Lipschitz uniformly semi-smooth* on V if $\forall \varepsilon > 0$
 $\exists \delta$ such that $\forall x, \forall y \in V$ with $\|x - y\| \leq \delta$,

$$\frac{\|F'(x; x - y) - F'(y; x - y)\|}{\|x - y\|} \leq \varepsilon$$

Proposition

$G : \mathbb{R}^m \rightarrow \mathbb{R}^n$, G uniformly semi-smooth* on V , then G is uniformly Newton differentiable on V with a Newton differential

$$\mathcal{H}(x) := \overline{\text{conv}} \left\{ H \in \mathbb{R}^{n \times m} \mid \exists \{x^k\}_{k \in \mathbb{N}}, \lim_{k \rightarrow \infty} \nabla G(x^k)^T = H \right\}$$

Non-smooth Newton

Given $G : \mathbb{R}^m \rightarrow \mathbb{R}^n$ Newton differentiable with $\mathcal{H}G$ Find, $\bar{x} \in \mathbb{R}^m$

$$G(\bar{x}) = 0 \quad (\text{UDE})$$

Denote $\mathcal{Z} = \{x \mid G(x) = 0\}$

Algorithm: $\mathcal{N} : \mathbb{R}^m \rightrightarrows \mathbb{R}^m$,

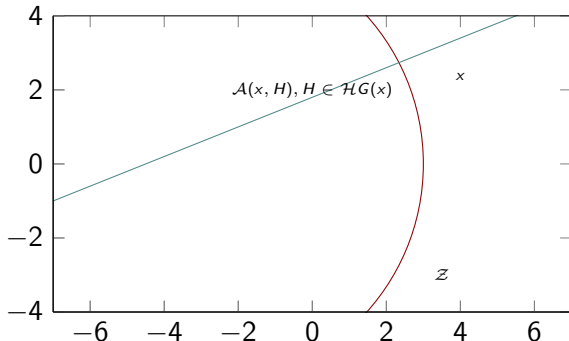
$$\mathcal{N}(x) = \{x - H^+ G(x) \mid H \in \mathcal{H}G(x)\}$$

Geometrically Compatible

Definition

$\mathcal{H}G$ is *geometrically compatible* if

$$\|P_{\mathcal{A}(x,H)}x - P_Z P_{\mathcal{A}(x,H)}x\| \leq P \|P_Z P_{\mathcal{A}(x,H)}x - P_{\mathcal{A}(x,H)}P_Z P_{\mathcal{A}(x,H)}x\|$$



If $\mathcal{H}G$ is single-valued and uniformly continuous then it is geometrically compatible

Convergence

Theorem

- ▶ $G : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is uniformly Newton differentiable on \mathcal{Z} with $\mathcal{H}G$ and Lipschitz
- ▶ $\mathcal{Z} := \{x \mid G(x) = 0\} \neq \emptyset$, $P_{\mathcal{Z}}$ is Lipschitz
- ▶ $\sup_{x \in U \in \mathcal{V}(\bar{x})} \sup_{H \in \mathcal{H}G(x)} \|H^+\| \leq \Omega$ and H full rank
- ▶ $\mathcal{H}G$ is geometrically compatible

Then the Newton-type method approaches \mathcal{Z} superlinearly for any x^0 close enough to \mathcal{Z}

$$\lim_{k \rightarrow \infty} \frac{\text{dist}(x^{k+1}, \mathcal{Z})}{\text{dist}(x^k, \mathcal{Z})} = 0$$

Constraints

Given $G : \mathbb{R}^m \rightarrow \mathbb{R}^n$ Newton differentiable with $\mathcal{H}G$ Find, $\bar{x} \in K$,
 K close and convex

$$G(\bar{x}) = 0 \quad (\text{CUDE})$$

Denote $\mathcal{Z} = \{x \mid G(x) = 0\}$, $\mathcal{K} = K \cap \mathcal{Z}$

Algorithm: $\mathcal{M} : \mathbb{R}^m \rightrightarrows \mathbb{R}^m$,

$$\mathcal{M} = P_K \circ \mathcal{M}$$

Using non-expansiveness of P_K , $y \in \mathcal{N}(x)$

$$\text{dist}(P_K y, \mathcal{K}) \leq \|P_K y - P_K x\| \leq \|y - P_K x\| \leq c \|x - P_K x\|$$

Back to Multi-objective

Problem

Find, $\bar{x} \in \mathbb{R}^m, \bar{\sigma} \in \Delta^n$

$$\bar{\sigma}^T \nabla F(\bar{x}) = 0 \quad (\text{FOE})$$

Denote $K = \mathbb{R}^m \times \Delta^n, \mathcal{K} = \{x, \sigma \mid \sigma \in \Delta^n, \sigma^T \nabla F(x) = 0\}$

Assume: ∇F is uniformly Newton differentiable on \mathcal{Z} with $\mathcal{H}F$

Algorithm: $H^k \in \mathcal{H}F(x^k)$

$$\begin{bmatrix} x^{k+1} \\ \sigma^{k+1} \end{bmatrix} = P_K \left(\begin{bmatrix} x^k \\ \sigma^k \end{bmatrix} - \begin{bmatrix} \sigma^k{}^T H^k & \nabla F(x^k) \end{bmatrix}^+ \sigma^k{}^T \nabla F(x^k) \right)$$

Main Result

Theorem

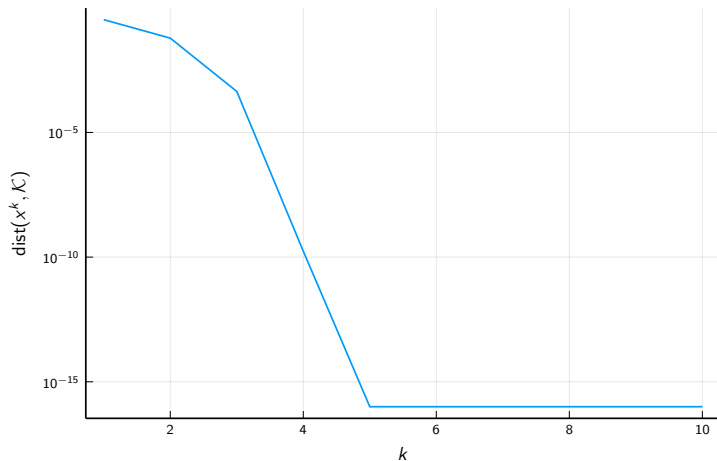
- ▶ $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is uniformly Newton differentiable on \mathcal{Z} with $\mathcal{H}F$ and Lipschitz
- ▶ $\mathcal{Z} := \{x, \sigma \mid \sigma^T \nabla F(x) = 0\} \neq \emptyset$, $P_{\mathcal{Z}}$ is Lipschitz
- ▶ $\sup_{x \in U \in \mathcal{V}(\bar{x})} \sup_{H \in \mathcal{H}G(x)} \|H^+\| \leq \Omega$ and H full rank
- ▶ $\mathcal{H}F$ is geometrically compatible

Then the Newton-type method approaches $\mathcal{Z} \cap \mathbb{R}^m \times \Delta^n$ superlinearly for any x^0 close enough to \mathcal{Z}

$$\lim_{k \rightarrow \infty} \frac{\text{dist}(x^{k+1}, \mathcal{Z} \cap \mathbb{R}^m \times \Delta^n)}{\text{dist}(x^k, \mathcal{Z} \cap \mathbb{R}^m \times \Delta^n)} = 0$$

Numerical Example

$$F(x) = [x_1^2 - x_2, x_2], \quad \mathcal{K} = \{0\} \times \mathbb{R} \times \{.5\} \times \{.5\}$$



Thank you!