

# Solution of Mismatched Monotone+Lipschitz Inclusion Problems

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Workshop DIPOpt

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## 1 Motivation

## 2 Existing Mismatch Algorithms

- Mismatched Forward-Backward Algorithm
- Mismatched Condat–Vũ Algorithm
- Mismatched Primal-Dual Algorithm

## 3 Main Problem and Results

- Mismatched Forward-Backward-Half-Forward
- Mismatched Forward-Douglas–Rachford-Forward
- Sketch of proof

## 4 Numerical Example

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## Optimization Problem

$$\min_{x \in \mathcal{H}} f(x) + \frac{\alpha}{2} \|Lx - c\|^2 \quad (1)$$

- $\mathcal{H}$  and  $\mathcal{G}$  are real Hilbert spaces.
- $f \in \Gamma_0(\mathcal{H})$ .
- $L : \mathcal{H} \rightarrow \mathcal{G}$  is a linear bounded operator.
- $\alpha \in ]0, +\infty[$ .

## Optimization Problem

$$\min_{x \in \mathcal{H}} f(x) + \frac{\alpha}{2} \|Lx - c\|^2 \quad (2)$$

Image restoration/reconstruction

- Recover an image  $x$  from the observation  $c$
- $L$ : observation process
- $f$ : regularization term
- $\frac{\alpha}{2} \|Lx - c\|^2$ : data fidelity term

## Optimization Problem

$$\min_{x \in \mathcal{H}} f(x) + \frac{\alpha}{2} \|Lx - c\|^2 \quad (3)$$

+  
(standard qualification conditions)



## Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in \partial f(x) + \alpha L^*(Lx - c). \quad (4)$$

## Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in \partial f(x) + \alpha L^*(Lx - c). \quad (5)$$

## Forward-Backward

$$(\forall n \in \mathbb{N}) \quad [z_{n+1} = \text{prox}_{\gamma f}(z_n - \gamma L^*(Lz_n - c))]. \quad (6)$$

$$\text{prox}_f : x \mapsto \arg \min_{y \in \mathcal{H}} \left( f(y) + \frac{1}{2} \|x - y\|^2 \right).$$

- Iterative methods require  $L^*$
- In some practical instances  $L^*$  is not available and it must be approximated

## Monotone Inclusion

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- Iterative methods require  $L^*$
- In some practical instances  $L^*$  is not available and it must be approximated

$$L^* \longrightarrow K$$



## Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in \partial f(x) + \alpha L^*(Lx - c). \quad (7)$$

$$L^* \longrightarrow K$$

## Adjoint Mismatch Problem

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in \partial f(x) + \alpha K(Lx - c). \quad (8)$$

- $K : \mathcal{G} \rightarrow \mathcal{H}$  is a linear bounded operator approximating  $L^*$ .
- The convergence results for classical splitting algorithms in the literature are no longer valid in this context.
- A solution to the adjoint mismatch problem is not necessarily a solution to the original problem.

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# Mismatched Forward-Backward (FB) Algorithm

## Adjoint Mismatch Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in \partial f(x) + K(Lx - c). \quad (9)$$

- $f \in \Gamma_0(\mathcal{H})$ .
- $L : \mathcal{H} \rightarrow \mathcal{G}$  and  $K : \mathcal{G} \rightarrow \mathcal{H}$  are linear bounded operators.

## Mismatched Forward-Backward Algorithm (Chouzenoux et al 2021)

$$(\forall n \in \mathbb{N}) \quad \left\{ \begin{array}{l} x_{n+1} = x_n + \theta_n (\text{prox}_{\gamma f}(x_n - \gamma K_n(Lx_n - c)) - x_n) \end{array} \right. \quad (10)$$

In the case when  $K_n \equiv L^*$  the classical Forward-Backward Algorithm is recovered.

## Remark

Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  and  $\beta \in ]0, +\infty[$ . The operator  $T$  is  $\beta$ -cocoercive if

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad \langle x - y \mid Tx - Ty \rangle \geq \beta \|Tx - Ty\|^2.$$

The operator  $T$  is  $\beta$ -Lipschitzian if

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad \|Tx - Ty\| \leq \beta \|x - y\|.$$

- Every cocoercive operator is a Lipschitzian operator, the converse is not true.
- If  $T = \nabla f$  where  $f$  is a convex differentiable function, the notion of Lipschitz and cocoercive gradients are equivalent (Baillon–Haddad).

## Mismatched Forward-Backward Algorithm (Chouzenoux et al 2021)

$$(\forall n \in \mathbb{N}) \quad \left[ x_{n+1} = x_n + \theta_n (\text{prox}_{\gamma f}(x_n - \gamma K_n(Lx_n - c)) - x_n) \right] \quad (11)$$

## Convergence (Chouzenoux et al 2021)

Suppose that

- 1  $\{K_n\}_{n \in \mathbb{N}} \subset \mathcal{B}(\mathcal{G}, \mathcal{H})$ , s.t.,  $\|K_n - K\| \leq \omega_n$ , where  $\sum_{n \in \mathbb{N}} \omega_n < +\infty$ .
- 2  $K \circ L$  is  $\eta$ -cocoercive.
- 3  $\gamma \in ]0, 2\eta[$ .
- 4  $\{\theta_n\}_{n \in \mathbb{N}} \subset [0, \delta]$ ,  $\sum \theta_n(\delta - \theta_n) = +\infty$ , where  $\delta = 2 - \gamma/(2\eta)$ .

Then  $x_n \rightarrow x \in \mathcal{H}$  such that  $x$  is a solution to (10).

## Cocoercivity of $K \circ L$ (Chouzenoux et al 2021)

Set  $T = K \circ L$ . Suppose that  $\text{ran}(T + T^*)$  is closed. Then,  $T$  is  $\eta$ -cocoercive if and only if  $\lambda_{\min} \geq 0$ ,  $\ker(T + T^*) = \ker(T)$ , and

$$\eta \leq \frac{2}{\|M\|^2}. \quad (12)$$

- $\lambda_{\min}$  is defined by

$$\lambda_{\min} = \inf \{ \langle x | KLx \rangle \mid x \in \mathcal{H}, \|x\| = 1 \}. \quad (13)$$

- $M = (\text{Id} + ((T - T^*)(T + T^*)^\#)(T + T^*)^{1/2})$
- $T^\#$  denotes the Moore-Penrose generalized inverse.

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# Mismatched Condat–Vũ (CV) Algorithm

## Adjoint Mismatch Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in \partial f(x) + N^* \partial g(Nx) + K(Lx - c). \quad (14)$$

- $g \in \Gamma_0(\mathcal{K})$ .
- $N : \mathcal{H} \rightarrow \mathcal{K}$  is a linear bounded operator.

## Mismatched Condat–Vũ Algorithm (Chouzenoux et al 2023)

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_{n+1} = \text{prox}_{\tau f}(x_n - \tau(K_n(Hx_n - c) + N^*u_n)) \\ v_{n+1} = \text{prox}_{\sigma \hat{g}^*}(u_n + \sigma N(2y_{n+1} - x_n)) \\ x_{n+1} = x_n + \theta_n(y_{n+1} - x_n) \\ u_{n+1} = u_n \theta_n(v_{n+1} - u_n) \end{cases} \quad (15)$$

In the case when  $K_n \equiv L^*$  the Condat–Vũ Algorithm is recovered.

## Mismatched Condat-Vũ Algorithm (Chouzenoux et al 2023)

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_{n+1} = \text{prox}_{\tau f}(x_n - \tau(K_n(Hx_n - c + N^*u_n))) \\ v_{n+1} = \text{prox}_{\sigma \hat{g}^*}(u_n + \sigma N(2y_{n+1} - x_n)) \\ x_{n+1} = x_n + \theta_n(y_{n+1} - x_n) \\ u_{n+1} = u_n \theta_n(v_{n+1} - u_n) \end{cases} \quad (16)$$

## Convergence (Chouzenoux et al 2023)

Suppose that

- 1  $\{K_n\}_{n \in \mathbb{N}} \subset \mathcal{B}(\mathcal{G}, \mathcal{H})$ , s.t.,  $\|K_n - K\| \leq \omega_n$ , where  $\sum_{n \in \mathbb{N}} \omega_n < +\infty$ .
- 2  $K \circ L$  is  $\eta$ -cocoercive.
- 3  $1 - \tau\delta\|N\|^2 > \tau\|M\|^2/4$
- 4  $\{\theta_n\}_{n \in \mathbb{N}} \subset [0, \delta]$ ,  $\sum \theta_n(\delta - \theta_n) = +\infty$ , where  $\delta = 2 - (1/\tau - \sigma\|N\|^2)^{-1}\|M\|^2/4$ .

Then  $x_n \rightarrow x \in \mathcal{H}$  such that  $x$  is a solution to (14).

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# Mismatched Primal-Dual (PD) Algorithm

## Adjoint Mismatch Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in \partial f(x) + K\partial h(Lx). \quad (17)$$

- $h \in \Gamma_0(\mathcal{G})$ .

## Mismatched Primal-Dual (Lorenz–Schnepp 2022)

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_{n+1} = \text{prox}_{\tau_n f}(x_n - \tau_n K u_n) \\ y_{n+1} = x_{n+1} + \lambda_n(x_{n+1} - x_n) \\ u_{n+1} = \text{prox}_{\sigma_n h^*}(u_n + \sigma_n L y_{n+1}) \end{cases} \quad (18)$$

In the case when  $K = L^*$  the Primal-Dual Algorithm (Chambolle–Pock 2011) is recovered.

## Mismatched Primal-Dual (Lorenz–Schnepp 2022)

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_{n+1} = \text{prox}_{\tau f}(x_n - \tau K^* u_n) \\ y_{n+1} = x_{n+1} + \lambda(x_{n+1} - x_n) \\ u_{n+1} = \text{prox}_{\sigma h^*}(u_n + \sigma L y_{n+1}) \end{cases} \quad (19)$$

## Convergence (Lorenz–Schnepp 2022)

If  $f$  and  $h$  are strongly convex functions, for adequate step-sizes  $\tau$ ,  $\sigma$  and extrapolation parameter  $\lambda$ , the sequence  $(x_n)_{n \in \mathbb{N}}$  converges linearly to a solution to (17).

- The authors provide examples where the convergence of the Mismatched Primal-Dual algorithm fails to converge if one of the functions is not strongly convex.

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# Main Problem

We will focus on the adjoint mismatch problem associated to the following inclusion

## Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + \alpha L^*Lx + L^*BLx, \quad (20)$$

where

- $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  is maximally monotone
- $C$  is  $\beta$ -cocoercive
- $B$  is monotone and  $\zeta$ -Lipschitzian

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- $B$  is monotone and  $\zeta$ -Lipschitzian

Example:

- $A = \partial f$
- $C = \nabla g$
- $B = \nabla h$



# Main Problem

Denote:  $D = \alpha L^*L + L^*BL$ .

## Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + Dx, \quad (21)$$

Note that  $D$  is monotone and Lipschitzian

$$\|L^*(\alpha \text{Id} + B)Lx - L^*(\alpha \text{Id} + B)Ly\| \leq \|L\|^2(\alpha + \zeta)\|x - y\|. \quad (22)$$

# Main Problem

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Therefore, the problem can be solved by:

- (FBHF) *Forward-Backward-Half-Forward* (Briceño-Arias–Davis 2018)

# Main Problem

Denote:  $D = \alpha L^*L + L^*BL$ .

## Monotone Inclusion

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Therefore, the problem can be solved by:

- (FBHF) *Forward-Backward-Half-Forward* (Briceño-Arias–Davis 2018)
- (FDRF) *Forward-Douglas–Rachford-Forward* (Ryu–Vũ 2020)

# Forward-Backward-Half-Forward (FBHF)

## FBHF (Briceño-Arias–Davis 2018)

$$(\forall n \in \mathbb{N}) \quad \begin{cases} u_n = Dz_n \\ y_n = z_n - \gamma(C(z_n) + u_n) \\ x_n = J_{\gamma A}(y_n) \\ z_{n+1} = x_n + \gamma(u_n - Dx_n). \end{cases} \quad (23)$$

## Convergence (Briceño-Arias–Davis 2018)

If

$$\gamma \in \left] 0, \frac{4\beta}{1 + \sqrt{1 + 16\beta^2 \|L\|^4 (\alpha + \zeta)^2}} \right[ , \quad (24)$$

then,  $z_n \rightarrow x \in \mathcal{H}$  where  $x$  is a solution to the monotone inclusion.

- $J_{\gamma A} = (\text{Id} + \gamma A)^{-1}$ , if  $A = \partial f$ , then  $J_{\gamma A} = \text{prox}_{\gamma f}$ .

# Remark

FBHF combines Forward-Backward (FB) and Forward-Backward-Forward (FBF - Tseng 2000).

$$\begin{cases} u_n = Dz_n \\ y_n = z_n - \gamma(Cz_n + u_n) \\ x_n = J_{\gamma A}(y_n) \\ z_{n+1} = x_n + \gamma(u_n - Dx_n). \end{cases} \xrightarrow[\underline{C=0}]{} \underbrace{\begin{cases} y_n = J_{\gamma A}(z_n - \gamma Dz_n) \\ z_{n+1} = y_n + \gamma(Dz_n - Dy_n). \end{cases}}_{\text{FBF}}$$
$$\Downarrow D=0$$
$$\underbrace{\begin{cases} z_{n+1} = J_{\gamma A}(z_n - \gamma C(z_n)). \end{cases}}_{\text{FB}}$$

# Forward-Douglas–Rachford-Forward (FDRF)

## FDRF (Ryu–Vũ 2020)

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_n = J_{\gamma C} z_n \\ w_n = D x_n \\ y_n = J_{\gamma A}(2x_n - z_n - \gamma w_n) \\ z_{n+1} = z_n + y_n - x_n - \gamma(D y_n - w_n). \end{cases} \quad (25)$$

## Convergence

If

$$\gamma \in \left\{ \gamma \in ]0, +\infty[ \mid \|L\|^4 (\alpha + \zeta)^2 \gamma^2 \left(1 + \frac{\gamma}{2\beta}\right) < 1 \right\}. \quad (26)$$

then,  $z_n \rightarrow x \in \mathcal{H}$  where  $J_{\gamma C} x$  is a solution to the monotone inclusion.

# Remark

FDRF combines Forward-Backward-Forward and Douglas–Rachford Splitting (DR - Lions–Mercier 1979).

$$\begin{array}{l} \left\{ \begin{array}{l} x_n = J_{\gamma C} z_n \\ w_n = D x_n \\ y_n = J_{\gamma A}(2x_n - z_n - \gamma w_n) \\ z_{n+1} = z_n + y_n - x_n - \gamma(Dy_n - w_n) \end{array} \right. \xrightarrow[C=0]{} \underbrace{\left\{ \begin{array}{l} y_n = J_{\gamma A}(z_n - \gamma D z_n) \\ z_{n+1} = y_n + \gamma(D z_n - D y_n) \end{array} \right.}_{\text{FBF}} \\ \quad \quad \quad \Downarrow D=0 \\ \underbrace{\left\{ z_{n+1} = J_{\gamma A}(2J_{\gamma C} z_n - z_n) + (\text{Id} - J_{\gamma C}) z_n \right\}}_{\text{DR}} \end{array}$$

# Mismatch Problem

## Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + \alpha L^*Lx + L^*BLx, \quad (27)$$

Consider the following relaxed monotone inclusion

## Adjoint Mismatch Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + \alpha K Lx + K B Lx \quad (28)$$

- $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  is a maximally  $\rho$ -monotone operator for some  $\rho \in \mathbb{R}$ , i.e., for every  $(x, u) \in \text{gra } A$  and  $(y, v) \in \text{gra } A$  we have

$$\langle x - y \mid u - v \rangle \geq \rho \|x - y\|^2. \quad (29)$$

- If  $\rho > -1$ ,  $J_A = (\text{Id} + A)^{-1}$  is single valued and Lipschitzian.
- We assume that the solution set is not empty.



## Adjoint Mismatch Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + \alpha KLx + KBLx \quad (30)$$

- $K(\alpha Id + B)L$  is Lipschitzian but not necessarily monotone.
- The convergence results for FBHF and FDRF are no longer valid in this context.

## Adjoint Mismatch Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + \alpha K Lx + K B Lx \quad (30)$$

- $K(\alpha Id + B)L$  is Lipschitzian but not necessarily monotone.
- The convergence results for FBHF and FDRF are no longer valid in this context.

Goals:

- Characterize the Monotone+Lipschitzian mismatch
- Extend the convergence of FBHF and FDRF for solving the mismatch problem allowing approximation on the iterations
- Relax hypotheses in previous results

## Notation

- $D_K: x \mapsto \alpha KLx + KBLx$
- $D_{K_n}: x \mapsto \alpha K_n Lx + K_n BLx$
- $\kappa_K$  be the Lipschitz constant of  $D_K$
- $\zeta_K$  be the Lipschitz constant of  $(K - L^*)BL$

$$\lambda_{\min} = \inf \{ \langle x | KLx \rangle \mid x \in \mathcal{H}, \|x\| = 1 \}. \quad (31)$$

## Adjoint Mismatch Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + D_K x. \quad (32)$$

## Proposition

Assume that  $\rho + \alpha\lambda_{\min} > 0$ . Then, there exists a unique solution  $z^*$  to the monotone inclusion (with  $L^*$ ). Furthermore, every solution  $z$  to the adjoint mismatch monotone inclusion is such that

$$\|z - z^*\| \leq \frac{1}{\rho + \alpha\lambda_{\min}} \|L^* - K\| \|(\alpha\text{Id} + BL)z\|. \quad (33)$$

- $\rho + \alpha\lambda_{\min} > 0$  is associated to the strongly monotonicity of  $A + D$ .

# Assumptions

We introduce the following assumptions:

## Assumption

1.  $\hat{\rho} = \rho + \alpha\lambda_{\min} - \zeta_K \geq 0$
2.  $(K_n)_{n \in \mathbb{N}}$  is a sequence of  $\mathcal{B}(\mathcal{G}, \mathcal{H})$  such that, for every  $n \in \mathbb{N}$ ,  $\|K_n - K\| \leq \omega_n$ , where  $\{\omega_n\}_{n \in \mathbb{N}} \subset [0, +\infty[$  and  $\sum_{n \in \mathbb{N}} \omega_n < +\infty$ .

## Remark

- Note that  $\zeta_K \leq \|L^* - K\| \|L\| \zeta$ .
- 1 guarantees the monotonicity of  $A + D_K$ .
- If  $B = 0$ . 1:  $\rho + \alpha\lambda_{\min} \geq 0$ .
- If  $\alpha = 0$ . 1:  $\rho - \zeta_K \geq 0 \Rightarrow \rho > 0$ .

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## Mismatched Forward-Backward-Half-Forward

$$(\forall n \in \mathbb{N}) \quad \begin{cases} u_n = D_{K_n} z_n \\ y_n = z_n - \gamma_n (C z_n + u_n) \\ x_n = J_{\gamma_n A}(y_n) \\ z_{n+1} = x_n + \gamma_n (u_n - D_{K_n} x_n). \end{cases} \quad (34)$$

### Theorem

Let  $(\gamma_n)_{n \in \mathbb{N}}$  be a sequence in  $[\varepsilon, \chi - \varepsilon]$ , for some  $\varepsilon \in ]0, \chi/2[$ , such that  $\gamma_n \rho > -1$ , where  $\chi = 4\beta / (1 + \sqrt{1 + 16\beta^2 \kappa_K^2})$  if  $\rho \geq 0$  and  $\chi = \min \left\{ 4\beta / (1 + \sqrt{1 + 16\beta^2 \kappa_K^2}), -1/\rho \right\}$  otherwise. Therefore

- 1  $(z_n)_{n \in \mathbb{N}}$  converges weakly to a zero of  $(A + C + D_K)$ .
- 2 If  $\hat{\rho} > 0$  and there exists  $\bar{\eta} \in [0, 1[$  such that, for every  $n \in \mathbb{N}$ ,  $\omega_n = \omega_0 \bar{\eta}^n$ , then  $(z_n)_{n \in \mathbb{N}}$  converges linearly to the unique zero of  $(A + C + D_K)$ .

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## Mismatched Forward-Douglas–Rachford-Forward

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_n = J_{\gamma C} z_n \\ w_n = D_{K_n} x_n \\ y_n = J_{\gamma A} (2x_n - z_n - \gamma w_n) \\ z_{n+1} = z_n + y_n - x_n - \gamma (D_{K_n} y_n - w_n). \end{cases} \quad (35)$$

### Theorem

Suppose that  $\gamma \in \left\{ \gamma \in ]0, +\infty[ \mid \kappa_K^2 \gamma^2 \left( 1 + \frac{\gamma}{2\beta} \right) < 1 \text{ and } \rho\gamma > -1 \right\}$ .

Then, the following hold.

- 1  $(z_n)_{n \in \mathbb{N}}$  converges weakly to some  $\bar{z} \in \mathcal{H}$  and  $(x_n)_{n \in \mathbb{N}}$  converges weakly to  $J_{\gamma C} \bar{z} \in \text{zer}(A + C + D_K)$ .
- 2 If  $\hat{\rho} > 0$  and there exists  $\bar{\eta} \in [0, 1[$  such that, for every  $n \in \mathbb{N}$ ,  $\omega_n = \omega_0 \bar{\eta}^n$ , then  $(z_n)_{n \in \mathbb{N}}$  converges linearly to  $\bar{z} \in \mathcal{H}$  and  $(x_n)_{n \in \mathbb{N}}$  converges linearly to  $J_{\gamma C} \bar{z}$  the unique zero of  $(A + C + D_K)$ .

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- Sketch of proof

## 4 Numerical Example

# Sketch of proof

Let  $Q^\gamma : \mathcal{H} \rightarrow \mathcal{H}$  be the operator associated to FBHF or FDRF ( $Q^\gamma = Q^\gamma(K)$ ). We can deduce the following Fejér type inequality.

$$(\forall z \in \mathcal{H})(\forall z^* \in \mathbf{Z}) \quad \|Q^\gamma z - z^*\|^2 \leq \|z - z^*\|^2 - \phi^\gamma(z), \quad (36)$$

where  $\phi^\gamma : \mathcal{H} \rightarrow [0, +\infty[$ .

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In order to consider approximations on  $K$  we consider an operator  $Q_n^\gamma : \mathcal{H} \rightarrow \mathcal{H}$  ( $Q_n^\gamma = Q_n^\gamma(K_n)$ ) approximating  $Q$  and we prove that

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where, for every  $z^* \in \mathbf{Z}$ ,

- $\{\xi_n(z^*)\}_{n \in \mathbb{N}} \subset [0, +\infty[$  such that  $\sum_{n \in \mathbb{N}} \xi_n(z^*) < +\infty$
- $\{\eta_n(z^*)\}_{n \in \mathbb{N}} \subset [0, +\infty[$  such that  $\sum_{n \in \mathbb{N}} \eta_n(z^*) < +\infty$ .

# Sketch of proof

Mismatched FBHF and Mismatched FDRF are defined by the recurrence:

$$(\forall n \in \mathbb{N}) \quad z_{n+1} = Q_n^\gamma z_n. \quad (38)$$

We can deduce the following quasi-Fejér inequality

$$\|z_{n+1} - z^*\| \leq (1 + \xi_n(z^*))\|z_n - z^*\| + \eta_n(z^*). \quad (39)$$

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- For linear convergence, a similar procedure is followed, using the strong monotonicity of  $A + D_K$ .

## Adjoint Mismatch Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Cx + \alpha K Lx + K B Lx \quad (41)$$

Mismatched Alg.	$A$	$C$	$B$	$\lambda_{\min}$
FB	$\mathcal{M}$	$0$	$0$	$\lambda_{\min} \geq 0$ (+ others)
CV	$\mathcal{M}$	$N^* \mathcal{M} N$	$0$	$\lambda_{\min} \geq 0$ (+ others)
CP	$\mathcal{S}$	$0$	$\mathcal{S}$	$\alpha = 0$
FBHF & FDRF	$\rho\text{-}\mathcal{M}$	$\mathcal{C}$	$\zeta\text{-}\mathcal{L}$	$\rho + \alpha \lambda_{\min} \geq \ L^* - K\  \ L\  \zeta$

- $\mathcal{M}$ : Maximally monotone
- $\mathcal{S}$ : Strongly monotone
- $\mathcal{C}$ : Cocoercive
- $\mathcal{L}$ : Lipschitzian

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## 4 Numerical Example

## Optimization Problem

$$\min_{x \in \mathbb{R}^N} f(x) + g(x) + \alpha \frac{\|Lx - c\|^2}{2} + h(Lx). \quad (42)$$

### Image reconstruction

- Recover an image  $x$  from the observation  $c$ .
- $f(x) = \iota_{[0, x_{\max}]^N}(x) + \frac{\rho}{2} \|x\|^2$
- $g(x) = (\Phi_\delta \circ W)(x)$  where  $W$  is an orthonormal wavelet transform and  $\Phi_\delta$  is the Huber function.
- $L \in \mathbb{R}^{N \times M}$  is a discretized Radon projector.
- $h$  is the generalized Anscombe function (smoothed approximation of the neg-log-likelihood associated to a Gauss-Poisson noise model).

# Numerical Example

## Optimization Problem

$$\min_{x \in \mathbb{R}^N} f(x) + g(x) + \alpha \frac{\|Lx - c\|^2}{2} + h(Lx). \quad (43)$$

↓

## Monotone Inclusion

$$\text{find } x \in \mathcal{H} \text{ such that } \partial f(x) + \partial g(x) + \alpha L^*(Lx - c) + L^* \nabla h(Lx). \quad (44)$$

↓


## Mismatch Monotone Inclusion

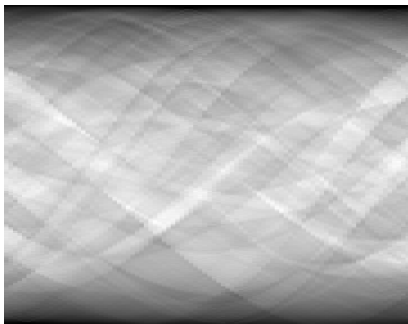
$$\text{find } x \in \mathcal{H} \text{ such that } \partial f(x) + \partial g(x) + \alpha K(Lx - c) + K \nabla h(Lx). \quad (45)$$

- $L$ : line length ray-drive projector (ASTRA<sup>1</sup> toolbox in MATLAB). 180°, 90 regularly spaced angular steps. Distance source to object 800 mm, source-to-image distance is 1200 mm.
- $K$ : adjoint of the strip fan-beam projector.
- $\lambda_{\min} \approx -6.0082$ . We set  $\hat{\rho} = -\alpha\lambda_{\min} + \hat{\zeta}_{L^*-K} + 10^{-3}$ .
- $\bar{x}$  represents a part of  $128 \times 128$  pixels of a high resolution scan of a phase-separated barium borosilicate glass imaged at the ESRF synchrotron<sup>2</sup>.
- $c = z + e$ , where  $z$  is a Poisson distribution with mean  $L\bar{x}$ , and  $e$  is a Gaussian distribution with zero-mean and variance  $\sigma^2$  ( $\sigma = 200$ ).
- $\lambda = 150$ ,  $\delta = 5$ , and  $\alpha = 0.1$ .

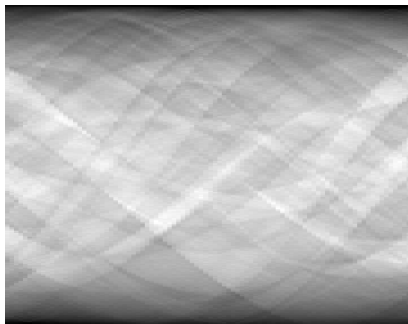
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<sup>1</sup>W. van Aarle, W.J. Palenstijn, J. De Beenhouwer, T. Altantzis, S. Bals, K.J. Batenburg, J. Sijbers: The astra toolbox: A platform for advanced algorithm development in electron tomography. Ultramicroscopy **157**, 35–47 (2015). <https://doi.org/10.1016/j.ultramic.2015.05.002>

<sup>2</sup><https://www.esrf.fr/> - The dataset is a courtesy of David Bouttes. 

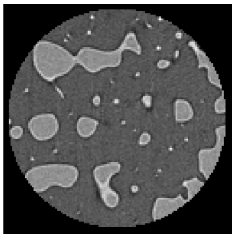


(a) Tomographic projection  $L\bar{x}$

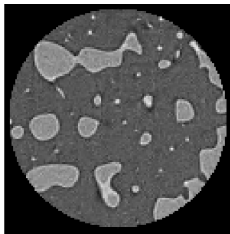


(b) Noisy projection  $c$ .

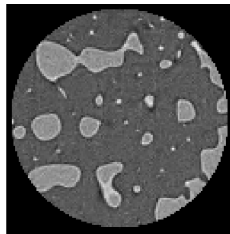
Figure: Clean projection and its noisy version, with  $\text{SNR}_{\text{input}} = 42.18$  dB



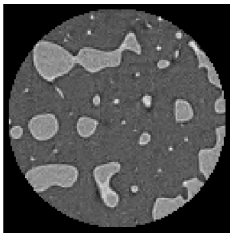
(a) Original image  $\bar{x}$



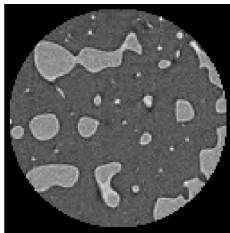
(b) MMFBHF SNR=23.25



(c) MMFDRF SNR=23.18



(d) FBHF SNR=23.31



(e) FDRF SNR=23.23

Figure: Original and reconstructed images after 1500 iterations.



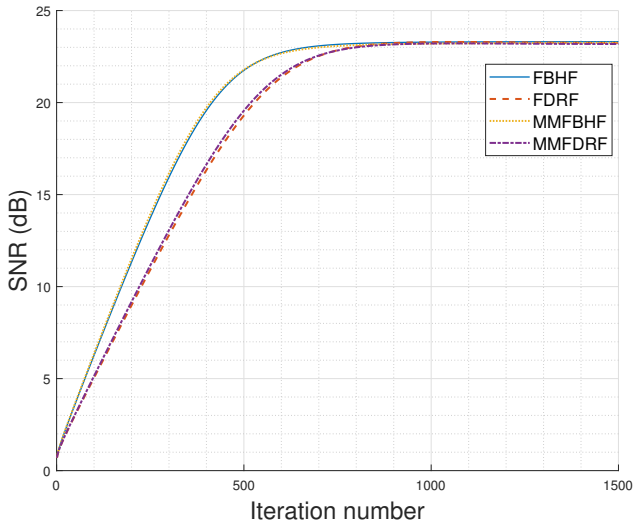


Figure: SNR (dB) evolution along iterations.



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# Thanks!