

5:18 PM

[illegible]

now the claims follow from the result below: /from Fermat's rule/

Proposition 11.4. (Parallel Scheduling Algorithm) Let $M = (M, \text{Dom}, P, L, \{1, \dots, n\}, \{1, \dots, m\}, S, \text{Dom})$.
Problem: $\exists \text{ dom } \xi \in L_1$ has atleast one solution, one of the following holds:
 $\forall i \in \{1, \dots, n\} \quad \text{dom } \xi_i = \emptyset$, $\forall i \in \{1, \dots, n\} \quad \text{dom } \xi_i = \{1\}$ or $\text{dom } \xi_i = \{1, 2\}$ or $\text{dom } \xi_i = \{1, 2, 3\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41\}$ or $\text{dom } \xi_i = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41,$

Proofs: Recall Proposition 25.7. Set $A_i := \partial f_i$ and recall $J_{A_i} = J_{A_i^*} = \text{Prox}_{A_i^*}$ for A_i maximally monotone.

(Corollary 2.9) (Forward-backward algorithm)

$f \in C_b(\mathbb{R}^n)$; $g: \mathbb{R}^n \rightarrow \mathbb{R}$, convex, differentiable, $\frac{1}{L}$ -Lipschitz-continuous gradient with $g \in \mathcal{R}_L$;

$\forall x \in \mathbb{D}, t \in \mathbb{R}$: $k = \min\{1, \frac{1}{L} + \frac{1}{2}\}$; $\|p_k\|_{\text{new}} \leq [0.5]$; $\lambda_k(\lambda_k - 1) = 0$; $x_k \neq 0$; $\text{Argmin}(f+g) \neq \emptyset$;

$\forall n$

$$\begin{cases} x_n = x_{n-1} \nabla g(x_{n-1}) \\ x_{n+1} = x_n + \lambda_n (\text{Prox}_{\frac{f}{\lambda_n}} x_n - x_n) \end{cases} \Rightarrow$$

(i) $\{x_n\}_{n \in \mathbb{N}}$: converges weakly to a point in $\text{Argmin}(f+g)$

(ii) $(\liminf_{n \rightarrow \infty} \lambda_n > 0 \wedge x \in \text{Argmin}(f+g)) \Rightarrow \nabla g(x) \rightarrow \nabla g(x)$

(iii) $(\liminf_{n \rightarrow \infty} \lambda_n > 0)$: one of the following holds:

(a) g : uniformly convex on every nonempty bounded subset of dom g

(b) g : uniformly convex on every nonempty bounded subset of $\mathcal{H} \Rightarrow$

$x \rightarrow$ unique minimizer of $f+g$

proof:

Set: $A = \{ \}$
 $R = \emptyset \Rightarrow A, B$: maximally ∇ subdifferential operator of a c. function is maximally monotone

now: dom $g = \mathbb{N}$ (for absurdum: $\exists x \in \mathbb{M} \quad g(x) = +\infty$)
 $\Rightarrow \text{int dom } g = \mathbb{N}$
 $\hookrightarrow g: \text{cannot be differentiable at } \tilde{x} \neq 1$

now: dom $g = M$ is per absurdum: $\exists \tilde{x} \in M$ $g(\tilde{x}) = \pm \infty$
 $\Rightarrow \text{int dom } g = M$ $\hookrightarrow g$: cannot be differentiable at $\tilde{x} \in M$
 $\therefore \text{dom}(g) \cap \text{int}(\text{dom}(g)) \neq \emptyset$
 as $\text{conv}(M)$ is $\text{conv}(M)$ as $\text{conv}(M)$

[illegible][illegible]

now:

- # Corollary 11.18 (Bolton-Muscatelli)
- (1) $\gamma \in \mathbb{R}$, Fréchet differentiable, convex, $B(\mathbb{R}^n, \mathbb{R})$
- $\gamma \in B$ -Lipschitz continuous $\Leftrightarrow \gamma \in \frac{1}{B}$ -COOPERATIVE
- $\gamma \in$ nonexpansive $\Leftrightarrow \gamma \in \frac{1}{B}$ -firmly nonexpansive

$B = \gamma \in B$ -COCOERCIVE

Finally we use the Theorem 25.2, which is the forward-backward algorithm in it's purest form:

Theorem 25.2 (Forward-backward algorithm) **JB: Strongly nonexpansive**

[A]: T -maximally monotone; $B \in \mathcal{H}_1$; $0, \eta, h, \beta$ -coercive; $\forall x \in Dg, \|x\| \leq \min\{\frac{\eta}{\beta}, \frac{1}{\beta} + \frac{1}{\beta}\}$

p_{n+1}^{FB} : sequence in (\mathcal{H}_1) ; $\sum_{n=0}^{\infty} \lambda_n (1-\alpha_n) = +\infty$; $x \in \mathcal{H}$; $\text{zer}(A) \neq \emptyset$;

$$y_n = x_n - \eta \nabla f(x_n) - \gamma \nabla \psi(x_n)$$

$$\begin{cases} x_{n+1} = x_n + \lambda_n (\frac{1}{\beta} y_n - x_n) \\ x_n = x_n + \lambda_n (\frac{1}{\beta} y_n - x_n) \end{cases} \Rightarrow x_n = x_n + \lambda_n (\frac{1}{\beta} y_n - x_n)$$

(i) $(p_n)_{n \in \mathbb{N}}$: converges weakly to a point in $\text{zer}(A) \cap B$

(ii) $(\inf_n \lambda_n > 0, x \in \text{zer}(A) \cap B) \Rightarrow (B p_n)_{n \in \mathbb{N}}: B x_n \rightarrow Bx$

(iii) $(\inf_n \lambda_n > 0)$: one of the following holds:

\Rightarrow (a) A : uniformly monotone on every nonempty bounded subset of $\text{dom} A$

(b) β : uniformly monotone on every nonempty bounded subset of \mathcal{H} \Rightarrow

(c) $\text{zer}(A)$: converges strongly to the unique point in $\text{zer}(A) \cap B$

Example 21.5:

$$[S \in L_2(A), Y \in R_H] \Rightarrow \begin{cases} \text{Pr}(Y_1 = 25) \\ \text{Pr}(Y_1 = 25) \end{cases}$$

```

[ 5: M=30, 400], prop, other, uniformly, sample, 5, dim, 2]
35: uniformly, sample, 5, dim, 2]

```

Primal problem: $\min_{x \in \mathcal{H}} Q(x) + \psi((x-r)) + \frac{1}{2} \|x-z\|^2$ [eq: 27.32]

Dual problem:

$$\min_{p^*, v^*} (p^*)^T (L^T v^* + z) + y^*{}^T (-v^*) - \langle v^*, r \rangle \quad [\text{eq: 27.33}]$$
$$\forall \epsilon \in]0, \frac{2}{\|L\|^2}[, \delta = \min \{1, \frac{\epsilon}{Y} \cdot \frac{1}{\|L\|^2}\} + \frac{1}{\epsilon}; (\lambda_n)_{n \in \mathbb{N}} \subseteq]0, \delta[, \inf_{n \in \mathbb{N}} \lambda_n > 0, \sup_{n \in \mathbb{N}} \lambda_n < \delta; \forall v \in E:$$
$$\begin{aligned} \text{Set: } & \quad V_{\text{new}} \quad \left| \quad \begin{aligned} & X_n = \text{Prox}_\varphi(L^T V_n + \tilde{z}) \\ & V_{n+1} = V_n - \lambda_n (\text{Prox}_{\omega\varphi}(\gamma(LX_n - Y) - V_n) + V_n) \end{aligned} \right. \quad [\text{eq: 27.34}] \end{aligned}$$

\bar{x} : unique solution to the primal problem \Rightarrow

(i) $V_n \rightarrow \tilde{V}$: solution to
 $\tilde{x} = \text{prox}_{\tilde{V}}(L^T \tilde{V} + z)$

$$(ii) \quad x_n \rightarrow \bar{x}$$

Proof: \mathbb{Z} : fixed vector

Set $h: \mathbb{H} \rightarrow]-\infty, +\infty]: x \mapsto \varphi(x) + \frac{1}{2} \|x - \tilde{z}\|^2 \in \Gamma_0(\mathbb{H}) \Rightarrow \text{dom } \varphi = \text{dom } h$
 $j: K \rightarrow]-\infty, +\infty]: y \mapsto \psi(y - r) \in \Gamma_0(K) \Rightarrow \text{dom } j = \text{dom } \psi$
 $\therefore j(y) = \psi(y - r) \Leftrightarrow \psi(\tilde{y}) = j(\tilde{y} + r)$

then primal problem : $\left(\min_{x \in \mathcal{H}} \underbrace{Q(x) + \frac{1}{2} \|x - z\|^2}_{h(x)} + \underbrace{\psi(Lx+r)}_{j(Lx)} \right) = \left(\min_{x \in \mathcal{H}} h(x) + j(Lx) \right) = \min (h + j \circ L)(H)$ [eq. 27.35]

given $x \in \text{ri}(L(\text{dom } \phi) - \text{dom } \psi)$ $\nexists \tilde{x} \in \text{ri } C \leftrightarrow \text{cone}(C - \tilde{x}) = \overline{\text{span}}(C - x) \neq \emptyset$

$$\Leftrightarrow \text{cone}(L(\text{dom } \Phi) - \text{dom } \Psi - r) = \overline{\text{span}}(L(\text{dom } \Phi) - \text{dom } \Psi - r)$$

now $L(\text{dom } \varphi) = \text{dom } \psi = r$

$$= L\{x \in H \mid \phi(x) < +\infty\} - \{y \in K \mid \psi(y) < +\infty\} - r$$
$$= \{x \in X \mid \varphi(x) < \infty\} - \{y \in Y \mid \psi(y) < \infty\} - r$$
$$= \{ (x, y, r) \mid \phi(x) < +\infty, \psi(y) < +\infty \}$$
$$= \{LX \sim (y+r) \mid \underbrace{\Phi(x) + \frac{1}{2}\|x-z\|^2}_{h(x)} < +\infty, j(y+r) < +\infty\}$$
$$= |\{x \mid h(x) < +\infty\} - \{y+r \mid j(y+r) < +\infty\}|$$
$$= L\{\underbrace{x | h(x) < +\infty}_{\text{dom } h} - \underbrace{\{j | i(j) < +\infty\}}_{\text{dom } i} = L(\text{dom } h) - \text{dom } j$$

Let $\vec{v} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$, $\vec{w} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$, $\vec{z} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$; set $\text{dom } \vec{v} = \{0, 1\}$
 Then in \mathbb{R}^2 , real Hilbert space, $\vec{v}, \vec{w}, \vec{z} \in \mathbb{R}^2$, $\langle \cdot, \cdot \rangle_{\mathbb{R}^2}$ one of the following holds:
 (i) $\text{dom}(\vec{v}) = \{0, 1\}$
 (ii) \vec{v} : finite-dimensional, \vec{v} polyhedral, $\text{dom } \vec{v} \cap \mathbb{R}^2 = \{0, 1\}$
 (iii) \vec{w} : finite-dimensional, \vec{w} polyhedral, $\text{dom } \vec{w} \cap \mathbb{R}^2 = \{0, 1\}$

$$\begin{aligned} \Rightarrow \nabla'(\varphi^* L) &= \underbrace{L^*}_{L} \cdot \nabla'(\varphi^*) \cdot L \\ \Rightarrow \nabla'(\varphi^*(Lv)) &= L \nabla'(\varphi^*) L^* v \\ \Rightarrow \nabla'(\underbrace{\varphi^*(Lv+z)}_{g(v)}) &= L \nabla'(\underbrace{\varphi^*}_{g(v)}) (Lv+z) \end{aligned}$$

$$\therefore \nabla g(v) = L \nabla^T(\Phi^*) (L^* v + z) \quad [\text{eq: 27.37}]$$

$$\begin{aligned} \therefore \forall_{v \in K} \quad \forall_{\omega \in K} \quad \| \nabla g(v) - \nabla g(\omega) \| &= \| L \nabla^*(\tilde{q}^*) (L^*v + z) - L \nabla^*(\tilde{q}^*) (L^*\omega + z) \| \\ &= \| L \left(\nabla^*(\tilde{q}^*) (L^*v + z) - \nabla^*(\tilde{q}^*) (L^*\omega + z) \right) \| \quad // \text{ use the def of linear-operator norm } \| \cdot \| \\ &\leq \| L \| \cdot \underbrace{\| \nabla^*(\tilde{q}^*) (L^*v + z) - \nabla^*(\tilde{q}^*) (L^*\omega + z) \|}_{\substack{// \leq \| L^*v + z - L^*\omega - z \| = \| L^*v - L^*\omega \| \\ // \leq \| L^* \| \cdot \| v - \omega \|}} \quad // \text{ now } \quad \nabla^*(\tilde{q}^*) \stackrel{1}{\text{ Lipschitz continuous }} \\ &\leq \| L \| \cdot \underbrace{\| L^*v - L^*\omega \|}_{\substack{// \| L^*(v - \omega) \| \leq \| L^* \| \cdot \| v - \omega \| \\ // \leq \| L^* \| \cdot \| v - \omega \|}} \quad // \text{ using Fact 3.12} \quad \| L \| = \| L^* \| \\ &\leq \| L \|^2 \| v - \omega \| \end{aligned}$$

$$\therefore \nabla g(v) = L \underbrace{\nabla'(\Phi^k)}_{\text{}} (L^k v + z) = L \text{prox}_{\Phi} (L^k v + z) \quad [\text{eq: 24.37.2}]$$

leg: 27.34], leg: 27.37] and leg: 27.38]

$$\hookrightarrow x_n = \text{prox}_D(L^T v_n + z) \Rightarrow Lx_n = L \text{prox}_D(L^T v_n + z) = \nabla g_n(v_n)$$

$$\begin{aligned}
 \gamma_{\text{best}} \cdot V_n - \lambda_n (\text{prox}_{\gamma \text{ best}} (r + (V_{n-1} - V_n) \cdot \gamma_{\text{best}}) - V_n) &= \text{prox}_{\gamma \text{ best}} (X(X_{n-1} - V_n) - V_n) = \text{prox}_{\gamma \text{ best}} (r + (\gamma(\gamma(V_n) - r) - V_n)) = \text{prox}_{\gamma \text{ best}} (-V_n - \gamma \cdot \gamma(V_n) + r + V_n) \\
 &= V_n - \lambda_n (\text{prox}_{\gamma \text{ best}} (V_n - \gamma \cdot \gamma(V_n)) - V_n) \quad \text{using } \frac{1}{\gamma} \\
 &= V_n + \lambda_n (\text{prox}_{\gamma \text{ best}} (V_n - \gamma \cdot \gamma(V_n)) + V_n) \\
 &= \text{prox}_{\gamma \text{ best}} (V_n - \gamma \cdot \gamma(V_n) + \gamma \cdot V_n) \quad \text{using } \frac{1}{\gamma}
 \end{aligned}$$

$$\begin{aligned}
 -V_n &= \text{prox}_{\gamma \Psi^*} \left(-(V_n - \gamma \nabla \Phi(V_n) + \gamma r) \right) \\
 &= \text{prox}_{(\gamma \Psi^*)^*} \left(V_n - \gamma \nabla \Phi(V_n) + \gamma r \right) \quad \text{by using } \# \\
 &= \text{prox}_{(\gamma \Psi^*)^*} \left(V_n - \gamma \nabla \Phi(V_n) - (-\gamma r) \right) \\
 &= \text{prox}_{(\gamma \Psi^*)^*} \left(V_n - \gamma \nabla \Phi(V_n) \right) \quad \text{by using } \# \\
 &= \text{prox}_{\gamma \Psi^*} \left(-(-\gamma r) \right) \quad \text{by using } \# \\
 &= \gamma \left(\Psi^* \right)^{-1}(-\gamma r) \\
 &= \gamma \left(\Psi^* \right)^{-1}(-\gamma r) = \gamma \# \\
 &\quad \# // \text{by definition} \\
 &= -\text{prox}_{\gamma \Psi} \left(V_n - \gamma \nabla \Phi(V_n) \right) \quad \#
 \end{aligned}$$

notice that this is the underlying recursion of forward-backward algorithm:

[illegible]

(b) $\inf_{\lambda \in \Lambda} \lambda_1 > 0$; one of the following holds:

- (i) λ_1 is a simple eigenvalue of A and $\lambda_1 > 0$.
- (ii) λ_1 is a multiple eigenvalue of A and $\lambda_1 > 0$.

λ_1 is the principal eigenvalue of A .

Proposition 0-4: If $\|B\|_2(D_0, \text{vec}(X); D_0N; \text{vec}(X))$, $\text{vec}(X) \in \text{vec}(D_0N)$, $\text{vec}(X) \in \text{vec}(D_0N)$, $\text{vec}(X) \in \text{vec}(D_0N)$
 Consider the problem:

$$\min_{\text{vec}(X)} \|B\|_2(D_0, \text{vec}(X); D_0N; \text{vec}(X)) \quad (0-4) \rightarrow \text{this is}$$

 together with the constraint:

$$\begin{aligned} & \min_{\text{vec}(X)} \left\{ \|B\|_2(D_0, \text{vec}(X); D_0N; \text{vec}(X)) \right\} \quad (0-4) \\ & = \min_{\text{vec}(X)} \left\{ \|B\|_2(D_0, \text{vec}(X); D_0N; \text{vec}(X)) \right\} \quad (0-4) \end{aligned}$$

 The solution to (0-4) is (0-4) has unique solution. $\bar{X} = \text{vec}(D_0N)$

y_k follows from \hat{y}

y_k converges weakly to a point in $\text{Argmin}(f|g)$ if $\exists: \forall m \exists^p (V^m - V) \rightarrow 0$ and $\exists: \forall m \exists^p (V^m - V) \rightarrow 0$ } **central problem** solves solution of our dual problem is $\text{Argmin}(f|g)$

hence it is \hat{y}

$\therefore \hat{x} = \text{Argmin}_x (L^*(\hat{y}))$

(ii) In (i) we have shown that $\vec{v}_* = \vec{v} : \vec{v} \in \text{Argmin}(\{f+g\})$ defined in eq: dual problem and $\vec{x} = \text{prox}_{\lambda}(\vec{L}^T \vec{v} + \vec{z})$: unique solution to dual problem

set $P = \sup_{n \in \mathbb{N}} \|v_n - \tilde{v}\|$ for
 as $(n_n)_{n \in \mathbb{N}}$ bounded \leftarrow
 $P < \infty$

Lemma 2.38. $\star \star \star$
 $(x_n)_{n \in \mathbb{N}}$: sequence in H
 $(x_n)_{n \in \mathbb{N}}$: converges weakly $\Leftrightarrow (x_n)$

$$\|x_n - \tilde{x}\|^2 \leq \|\text{prox}_\lambda(x_n + z) - \text{prox}_\lambda(x^* + z)\|^2$$

$$\overset{\text{V}_{\text{NEN}}}{\leq} \underbrace{\|\text{prox}_\lambda(x_n^* + z) - \text{prox}_\lambda(x^* + z)\|}_{\text{by definition}}$$

$$\leq \langle L^T v_n + z^T / \lambda^2 - z^T / \lambda^2, \text{prox}_\lambda(x_n^* + z) - \text{prox}_\lambda(x^* + z) \rangle$$

$$= \langle L^T (V_n - \tilde{V}) \rangle \text{prox}_\lambda(x_n^* + z) - \text{prox}_\lambda(x^* + z)$$

$$\overset{\text{linear operator } A}{=} \langle V_n - \tilde{V} \rangle \text{prox}_\lambda(x_n^* + z) - \text{prox}_\lambda(x^* + z)$$

$$= \langle V_n - \tilde{V} \rangle L \cdot (\text{prox}_\lambda(x_n^* + z) - \text{prox}_\lambda(x^* + z))$$

$$\langle V_n - \tilde{V} \rangle L \text{prox}_\lambda(x_n^* + z) - L \text{prox}_\lambda(x^* + z)$$

$$\underbrace{\langle V_n - \tilde{V} \rangle L \text{prox}_\lambda(x_n^* + z)}_{\mathcal{V}_n(V_n)} - \underbrace{L \text{prox}_\lambda(x^* + z)}_{\mathcal{V}_n(\tilde{V})}$$

$$\text{4. from } \text{lem. 24.2.2) : } \mathcal{V}_n(\tilde{V}) = L \text{prox}_\lambda(x^* + z) \quad \forall$$

$$\langle V_n - \tilde{V} \rangle \mathcal{V}_n(V_n) - \mathcal{V}_n(\tilde{V}) \leq \|V_n - \tilde{V}\| \cdot \|\mathcal{V}_n(V_n) - \mathcal{V}_n(\tilde{V})\|$$

$$\overset{\text{V}_{\text{NEN}}}{\leq} \underbrace{\left(\sup_{\text{NEN}} \|V_n - \tilde{V}\| \right)}_{\text{finite, positive}} \|\mathcal{V}_n(V_n) - \mathcal{V}_n(\tilde{V})\|$$

$$\therefore \underbrace{\|x_n - \tilde{x}\|^2}_{\text{V}_{\text{NEN}}} \leq \underbrace{C}_{0 \text{ as } n \rightarrow \infty} \|\mathcal{V}_n(V_n) - \mathcal{V}_n(\tilde{V})\|$$

$$\therefore \|x_n - \bar{x}\| \rightarrow 0 \text{ as } n \rightarrow \infty$$
$$\Leftrightarrow \boxed{x_n \rightarrow x} \quad \checkmark$$

If these are formulas for integral operators for functions of importance

PROPOSITION 11.23 (Properties of probability operators. Some results are)

[56] (M1), (M2), (VER₁) [] where $\forall x, y \in \mathcal{M}$ then $\text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(i) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall w \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(ii) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(iii) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(iv) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(v) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(vi) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(vii) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$

(viii) $\forall x, y \in \mathcal{M}, \forall z \in \mathcal{M}, \forall u \in \mathcal{M}, \forall v \in \mathcal{M} \Rightarrow \text{Prob}_{\frac{1}{2}(\mathcal{M})} x = \text{Prob}_{\frac{1}{2}(\mathcal{M})} (x \vee y)$