

# When ExtraGradient meets VR: bridging two giants to boost VIs

part II

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# *In previous episodes*

*Variational inequality:*

$$\forall z \in \mathbb{Z} \rightarrow \langle F(z^*), z - z^* \rangle \geq 0$$

*Stochastic setting:*

$$F(z) = \frac{1}{n} \sum_{i=1}^n F_i(z)$$

*Dispersion problem:*

$$\mathbb{E}[\|F_\xi(z^t) - F(z^t)\|^2] \not\rightarrow 0 \quad \text{with condition} \quad z^t \rightarrow z^*$$

# ExtraSAGA

$$z^{t+\frac{1}{2}} = z^t - \gamma \frac{1}{n} \sum_{i=1}^n y_i^t$$

$$G^t = F_{i_t}(z^{t+\frac{1}{2}}) - y_{i_t}^t + \frac{1}{n} \sum_{i=1}^n y_i^t$$

$$y_i^{t+1} = F_{i_t}(z^{t+\frac{1}{2}}) \quad for \quad i = i_t$$

$$z^{t+1} = z^t - \gamma G^t$$

# Convergence with the proper step 1/6

$$\begin{aligned}
 \|z^{t+1} - z^*\|^2 &\leq \left(1 - \frac{\gamma\mu}{2}\right) \|z^t - z^*\|^2 - (1 - \gamma\mu) \|z^{t+\frac{1}{2}} - z^t\|^2 + && \text{Descent Lemma} \\
 &\quad + 2\gamma \left\langle F(z^{t+\frac{1}{2}}) - G^t, z^{t+\frac{1}{2}} - z^* \right\rangle - \gamma \left\langle F(z^{t+\frac{1}{2}}) - F(z^*), z^{t+\frac{1}{2}} - z^* \right\rangle + \\
 &\quad + \gamma^2 \left\| \frac{1}{n} \sum_j y_j^t - G^t \right\|^2
 \end{aligned}$$

Variance Reduction

$$\begin{aligned}
 \mathbb{E} \left\| \frac{1}{n} \sum_j y_j^t - G^t \right\|^2 &= \mathbb{E} \left\| \frac{1}{n} \sum_j y_j^t - F_{i_t}(z^{t+\frac{1}{2}}) + y_{i_t}^t - \frac{1}{n} \sum_j y_j^t \right\|^2 \\
 &= \mathbb{E} \left\| y_{i_t}^t - F_{i_t}(z^*) + F_{i_t}(z^*) - F_{i_t}(z^{t+\frac{1}{2}}) \right\|^2 \\
 &\leq 2\mathbb{E} \|y_{i_t}^t - F_{i_t}(z^*)\|^2 + 2\mathbb{E} \|F_{i_t}(z^{t+\frac{1}{2}}) - F_{i_t}(z^*)\|^2 \\
 &\leq \underbrace{\frac{2}{n} \sum_j \|y_j^t - F_j(z^*)\|^2}_{\sigma_t^2} + 2L \left\langle F(z^{t+\frac{1}{2}}) - F(z^*), z^{t+\frac{1}{2}} - z^* \right\rangle
 \end{aligned}$$

# Final view of linear convergence in a case of mu-strong monotone operator

$$\mathbb{E}[\|z^{t+1} - z^*\| + M\sigma_{t+1}^2 |z^{t+\frac{1}{2}}|] \leq \left(1 - \frac{\gamma\mu}{2}\right) \|z^t - z^*\|^2 + \left(1 - \frac{1}{2n}\right) \sigma_t^2$$

## Monotone operator case

$$\text{Gap}(w) = \max_{z \in C} \{ \langle F(z), w - z \rangle \} \quad \text{Special technic}$$

$$\mathbb{E} [\text{Gap}(z^K)] = O\left(\frac{L}{K}\right) \quad \text{for} \quad z^K = \frac{1}{K} \sum_{i=0}^{K-1} z^{i+\frac{1}{2}}$$

$$\gamma K \mathbb{E} [\text{Gap}(z^K)] \leq \max_{z \in C} \|z_0 - z\|^2 + \sigma_K^2$$

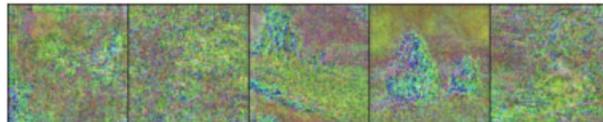
! Sublinear Convergence !

# Positive Result

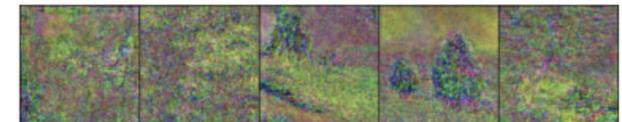
Real Images from Domain X



Generated Images to Domain Y



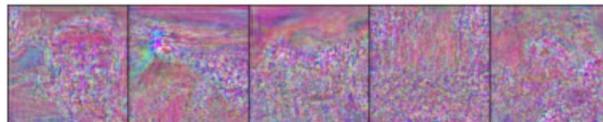
Generated Images to Domain Y



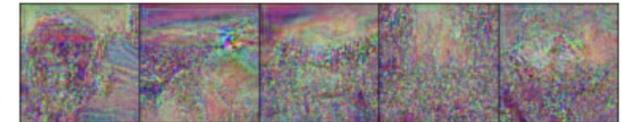
Real Images from Domain Y



Generated Images to Domain X



Generated Images to Domain X



Epoch 10

Epoch 160

Idea that coefficient makes ExtraSAGA effective

$$G^t = F_{i_t}(z^{t+\frac{1}{2}}) - y_{i_t}^t + \frac{1}{n} \sum_{i=1}^n y_i^t$$

Estimator

$$X^* = X - \alpha(Y - \mathbb{E}Y)$$

Variance through  
affection

Suggestion for minimizing

$$\alpha^* = \frac{\text{Cov}(X, Y)}{\text{Var}(Y)}$$

# Our intentions:

To make an explorations and experiments with alpha- method;

To achieve positive in Denoising;

GAN and Adversarial training;

Article writing;

Thanks for attention