SGD Ex

July 12, 2022

[P4]

E1

Def. 17:

$$D_f(x,y) + D_f(y,x) = \langle \nabla f(x) - \nabla f(y), x - y \rangle = \langle \nabla f(y) - \nabla f(x), y - x \rangle \tag{1}$$

 $\forall x, y \in \mathbb{R}^d$:

$$\mu||x-y||^{2} \leq 2D_{f}(x,y),
\frac{\mu}{2}||x-y||^{2} \leq D_{f}(x,y),
\frac{\mu}{2}||x-y||^{2} \leq D_{f}(y,x),
D_{f}(x,y) + \frac{\mu}{2}||x-y||^{2} \leq D_{f}(x,y) + D_{f}(y,x),
D_{f}(x,y) + \frac{\mu}{2}||x-y||^{2} \leq \langle \nabla f(x) - \nabla f(y), x-y \rangle.$$
(2)

E2

$$D_{f}(x,y) + \frac{\mu}{2}||x-y||^{2} \leq \langle \nabla f(x) - \nabla f(y), x-y \rangle,$$

$$\langle \nabla f(x) - \nabla f(y), x-y \rangle \geq \underbrace{D_{f}(x,y)}_{\geq \frac{\mu}{2}||x-y||^{2}} + \frac{\mu}{2}||x-y||^{2},$$

$$\langle \nabla f(x) - \nabla f(y), x-y \rangle \geq \frac{\mu}{2}||x-y||^{2} + \frac{\mu}{2}||x-y||^{2},$$

$$\langle \nabla f(x) - \nabla f(y), x-y \rangle \geq \mu||x-y||^{2}.$$
(3)

[P6]

E17

(Equation 34):

$$\langle a, b \rangle \leq \frac{||a||^2}{2t} + \frac{t||b||^2}{2},$$

$$\langle a, b \rangle \leq \frac{\langle a, a \rangle}{2t} + \frac{t\langle b, b \rangle}{2},$$

$$2t\langle a, b \rangle \leq \langle a, a \rangle + t^2\langle b, b \rangle,$$

$$0 \leq \langle a, a \rangle + \langle tb, tb \rangle - \langle a, tb \rangle - \langle tb, a \rangle,$$

$$0 \leq ||a - tb||^2.$$
(4)

(Equation 35):

$$||a+b||^{2} \leq 2||a||^{2} + 2||b||^{2},$$

$$\langle a, a \rangle + \langle b, b \rangle + 2\langle a, b \rangle \leq 2\langle a, a \rangle + 2\langle b, b \rangle,$$

$$0 \leq \langle a, a \rangle + \langle b, b \rangle - 2\langle a, b \rangle,$$

$$0 \leq ||a-b||^{2}.$$
(5)

(Equation 36):

$$\frac{1}{2}||a||^2 - ||b||^2 \le ||a+b||^2,$$

$$\frac{1}{2}\langle a, a \rangle - \langle a, a \rangle \le \langle a, a \rangle + \langle b, b \rangle + 2\langle a, b \rangle,$$

$$\langle a, a \rangle - 2\langle b, b \rangle \le 2\langle a, a \rangle + 2\langle b, b \rangle + 4\langle a, b \rangle,$$

$$0 \le \langle a, a \rangle + \langle 2b, 2b \rangle + \langle a, 2b \rangle + \langle 2b, a \rangle,$$

$$0 \le ||a+2b||^2.$$
(6)

E19

For random vector $X \in \mathbb{R}^d$:

$$\mathbf{Var}[X] := \mathbf{E}\left[||X - \mathbf{E}[X]||^2\right]. \tag{7}$$

Markov's inequality:

$$\operatorname{Prob}(X \ge t) \le \frac{\mathbf{E}[X]}{t}.\tag{8}$$

Proof of Chebyshev's inequality using Markov's inequality:

$$\text{Prob}(||X - \mathbf{E}[X]||^2 \ge t^2) \le \frac{\mathbf{E}[||X - \mathbf{E}[X]||^2]}{t^2}.$$

Since

$$\operatorname{Prob}(||X - \mathbf{E}[X]||^2 \ge t^2) = \operatorname{Prob}(||X - \mathbf{E}[X]|| \ge t), \tag{9}$$

then

$$\operatorname{Prob}(||X - \mathbf{E}[X]|| \ge t) \le \frac{\operatorname{Var}[X]}{t^2}.$$
 (10)

[P7]

E24

If

$$f = \frac{1}{n} \sum_{i=1}^{n} f_i,$$

then

$$D_f(x,y) = \frac{1}{n} \sum_{i=1}^n f_i(x) - \frac{1}{n} \sum_{i=1}^n f_i(y) - \frac{1}{n} \sum_{i=1}^n \langle \nabla f_i(y), x - y \rangle,$$

$$D_f(x,y) = \frac{1}{n} \sum_{i=1}^n (f_i(x) - f_i(y) - \langle \nabla f_i(y), x - y \rangle),$$

$$D_f(x,y) = \frac{1}{n} \sum_{i=1}^n D_{f_i}(x,y).$$

E26

If $\sigma_{\star}^2 = 0$, then

$$\sigma_{\star}^{2} = \left(\frac{1}{n^{2}} \sum_{i=1}^{n} \frac{||\nabla f_{i}(x^{\star})||^{2}}{p_{i}}\right) - ||\nabla f(x^{\star})||^{2} = 0$$

$$= \left(\frac{1}{n^{2}} \sum_{i=1}^{n} \frac{||np_{i}\nabla f(x^{\star})||^{2}}{p_{i}}\right) - ||\nabla f(x^{\star})||^{2} = 0$$

$$= p_{i} \sum_{i=1}^{n} (||\nabla f(x^{\star})||^{2}) - ||\nabla f(x^{\star})||^{2} = 0$$

$$= np_{i} ||\nabla f(x^{\star})||^{2} - ||\nabla f(x^{\star})||^{2} = 0,$$

$$\sigma_{\star}^{2} = 0 \implies np_{i} \nabla f(x^{\star}) = \nabla f(x^{\star}).$$

[P8]

E33

Let

$$\chi_i = \begin{cases} 1 & i \in S \\ 0 & i \notin S \end{cases} .$$

Since

$$p_i = \frac{1}{n},$$

and

$$|S| = \tau,$$

then

$$\mathbf{E}[\chi_i] = \operatorname{Prob}(i \in S) = \sum_{i=1}^n p_i \chi_i = \frac{1}{n} \sum_{i=1}^n \chi_i = \frac{\tau}{n}.$$

E35

For any vectors, $b_1, ..., b_n \in \mathbb{R}^d$:

$$\left\| \sum_{i=1}^{n} b_{i} \right\|^{2} - \sum_{i=1}^{n} \|b_{i}\|^{2} = \underbrace{\sum_{i=1}^{n} \langle b_{i}, b_{i} \rangle + \sum_{i \neq j} \langle b_{i}, b_{j} \rangle}_{\left\| \sum_{i=1}^{n} b_{i} \right\|^{2}} - \sum_{i=1}^{n} \|b_{i}\|^{2} = \underbrace{\sum_{i \neq j} \langle b_{i}, b_{j} \rangle}_{i \neq j}.$$

[P9]

E37

Assumptions of $C: \mathbb{R}^d \to \mathbb{R}^d$:

1.
$$\mathbf{E}[\mathcal{C}(x)] = x, \quad \forall x \in \mathbb{R}^d$$

2.
$$\mathbf{E}[||\mathcal{C}(x) - x||^2] \le \omega ||x||^2 + \delta, \quad \forall x \in \mathbb{R}^d, \quad \exists \omega, \delta \ge 0$$

Proof of convergence for CGD with n = 1: Since $C \in \mathbb{B}^d(\omega)$,

$$\mathbf{E}\left[||g(x)||^2\right] = \mathbf{E}\left[||\mathcal{C}(\nabla f(x))||^2\right] \le (\omega + 1)||\nabla f(x)||^2. \tag{11}$$

In case of $\nabla f(y) = 0$,

$$G(x,y) := \mathbf{E} \left[||g(x) - \nabla f(y)||^2 \right]$$

$$= \mathbf{E} \left[||g(x)||^2 \right]$$

$$\stackrel{\text{(11)}}{\leq} (\omega + 1) ||\nabla f(x) - \nabla f(y)||^2,$$

$$\stackrel{\text{(22)}}{\leq} 2(\omega + 1) LD_f(x, y).$$

In case of $\nabla f(y) \neq 0$,

$$G(x,y) := \mathbf{E} \left[||g(x) - \nabla f(y)||^{2} \right]$$

$$= \mathbf{E} \left[||g(x) - \nabla f(x)||^{2} \right] + ||\nabla f(x) - \nabla f(y)||^{2}$$

$$= \mathbf{E} \left[||\mathcal{C}(\nabla f(x)) - \nabla f(x)||^{2} \right] + ||\nabla f(x) - \nabla f(y)||^{2}$$

$$\leq \omega ||\nabla f(x)||^{2} + \delta + ||\nabla f(x) - \nabla f(y)||^{2}$$

$$= \omega ||\nabla f(x) - \nabla f(y) + \nabla f(y)||^{2} + ||\nabla f(x) - \nabla f(y)||^{2} + \delta$$

$$\leq 2\omega ||\nabla f(x) - \nabla f(y)||^{2} + 2\omega ||\nabla f(y)||^{2} + ||\nabla f(x) - \nabla f(y)||^{2} + \delta$$

$$= (2\omega + 1)||\nabla f(x) - \nabla f(y)||^{2} + 2\omega ||\nabla f(y)||^{2} + \delta$$

$$\leq 2\underbrace{(2\omega + 1)L}_{A} D_{f}(x, y) + \underbrace{2\omega ||\nabla f(y)||^{2} + \delta}_{C}.$$

If $0 < \gamma < \frac{1}{A}$, then

$$\mathbf{E}\left[||x^k - x^*||^2\right] \le (1 - \gamma\mu)^k ||x^0 - x^*|| + \frac{2\gamma\omega||\nabla f(x^*)||^2 + \gamma\delta}{\mu}.$$

E39

Lemma 51:

if $C(x) = x, \forall x$ (no master compression) and $\omega_i = \omega, \forall i$, then

$$G(x,y) \le 2 \underbrace{\left(L + 2L_{\max} \frac{\omega}{n}\right)}_{A} D_f(x,y) + \underbrace{2\frac{\omega}{n} \sigma^2(y)}_{C(y)},$$

where

$$\sigma^{2}(y) := \frac{1}{n} \sum_{i=1}^{n} ||\nabla f_{i}(y)||^{2}.$$

If $\sigma^2(y) = 0$, then

$$G(x,y) \le 2\underbrace{\left(L + L_{\max} \frac{\omega}{n}\right)}_{A} D_f(x,y).$$

Proof:

If $\nabla f(y) \neq 0$, then

$$G(x,y) := \mathbf{E} \left[||g(x) - \nabla f(y)||^2 \right]$$

$$= \mathbf{E} \left[||g(x) - \nabla f(x)||^2 \right] + ||\nabla f(x) - \nabla f(y)||^2$$

$$\leq \mathbf{E} \left[||g(x) - \nabla f(x)||^2 \right] + 2LD_f(x,y),$$
(12)

and

$$g(x) = \mathcal{C}(\hat{g}(x)) = \hat{g}(x) = \frac{1}{n} \sum_{i=1}^{n} g_i(x).$$
 (13)

where

$$g_i(x) = C_i(\nabla f_i(x)).$$

Estimate

$$\mathbf{E} \left[||g(x) - \nabla f(x)||^2 \right]^{\underbrace{13}}_{=} \mathbf{E} \left[||\mathcal{C}(\hat{g}(x)) - \nabla f(x)||^2 \right]$$

$$= \mathbf{E} \left[||\hat{g}(x) - \nabla f(x)||^2 \right]$$

$$= \mathbf{E} \left[\left\| \frac{1}{n} \sum_{i=1}^n \underbrace{(g_i(x) - \nabla f_i(x))}_{a_i} \right\|^2 \right]$$

$$= \frac{1}{n^2} \mathbf{E} \left[\sum_{i=1}^n ||a_i||^2 + \sum_{i \neq j} \langle a_i, a_j \rangle \right]$$

$$= \frac{1}{n^2} \sum_{i=1}^n \mathbf{E} \left[||a_i||^2 \right] + \sum_{i \neq j} \mathbf{E} \left[\langle a_i, a_j \rangle \right]$$

$$= \frac{1}{n^2} \sum_{i=1}^n \mathbf{E} \left[||a_i||^2 \right] + \sum_{i \neq j} \langle \underbrace{\mathbf{E} \left[a_i \right]}_{0}, \underbrace{\mathbf{E} \left[a_j \right]}_{0} \rangle$$

$$\leq \frac{1}{n^2} \sum_{i=1}^n \omega_i ||\nabla f_i(x)||^2$$

$$= \frac{\omega}{n^2} \sum_{i=1}^n ||\nabla f_i(x)||^2.$$

Next, bound

$$||\nabla f_i(x)||^2 = ||\nabla f_i(x) - \nabla f_i(y) + \nabla f_i(y)||^2$$

$$\leq 2||\nabla f_i(x) - \nabla f_i(y)||^2 + 2||\nabla f_i(y)||^2$$

$$\leq 4L_i D_{f_i}(x, y) + 2||\nabla f_i(y)||^2.$$

Combine everything:

$$G(x,y) \leq \mathbf{E} \left[||g(x) - \nabla f(x)||^{2} \right] + 2LD_{f}(x,y)$$

$$\leq \frac{\omega}{n^{2}} \sum_{i=1}^{n} ||\nabla f_{i}(x)||^{2} + 2LD_{f}(x,y)$$

$$\leq \frac{\omega}{n^{2}} \sum_{i=1}^{n} \left(4L_{i}D_{f_{i}}(x,y) + 2||\nabla f_{i}(y)||^{2} \right) + 2LD_{f}(x,y)$$

$$= 2\frac{\omega}{n} \left(2\sum_{i=1}^{n} \frac{1}{n}L_{i}D_{f_{i}}(x,y) + \frac{1}{n}\sum_{i=1}^{n} ||\nabla f_{i}(y)||^{2} \right) + 2LD_{f}(x,y)$$

$$\leq 2\frac{\omega}{n} \left(2L_{\max}D_{f}(x,y) + \sigma^{2}(y) \right) + 2LD_{f}(x,y)$$

$$= 2(L + 2L_{\max})D_{f}(x,y) + 2\frac{\omega}{n}\sigma^{2}(y).$$
(14)

Else, if $\nabla f(y) = 0$, then

$$G(x,y) = \mathbf{E} \left[||\hat{g}(x)||^{2} \right]$$

$$= \mathbf{E} \left[||\hat{g}(x) - \mathbf{E} [\hat{g}(x)]||^{2} \right] + ||\mathbf{E} [\hat{g}(x)]||^{2}$$

$$= \mathbf{E} \left[||\hat{g}(x) - \nabla f(x)||^{2} \right] + ||\nabla f(x)||^{2}$$

$$\leq \left(\frac{\omega}{n^{2}} \sum_{i=1}^{n} ||\nabla f_{i}(x)||^{2} \right) + ||\nabla f(x) - \nabla f(y) + \nabla f(y)||^{2}$$

$$\leq \left(\frac{\omega}{n^{2}} \sum_{i=1}^{n} ||\nabla f_{i}(x)||^{2} \right) + 2||\nabla f(x) - \nabla f(y)||^{2} + 2||\nabla f(y)||^{2}$$

$$\leq \left(\frac{\omega}{n^{2}} \sum_{i=1}^{n} ||\nabla f_{i}(x)||^{2} \right) + 2LD_{f}(x,y)$$

$$\leq \dots \text{ same as } (14), \text{ from the third line}$$

$$= 2(L + 2L_{\text{max}})D_{f}(x,y) + 2\frac{\omega}{n}\sigma^{2}(y).$$
(15)

[P10]

E41

Let

$$p_i = \text{Prob}(i \in S),$$

where

$$S \subseteq \{1, 2, ..., d\},\$$

then

$$\mathbf{E}[|S|] = \mathbf{E}\left[\sum_{i=1}^{d} |S_i|\right] = \sum_{i=1}^{d} \mathbf{E}[|S_i|] = \sum_{i=1}^{d} 1p_i + 0(1 - p_i) = \sum_{i=1}^{d} p_i.$$

E42

If $\mathbf{E}[\mathbf{C}^{\top}\mathbf{C}]$ is finite, then $\forall x \neq 0$:

$$x^{T}\mathbf{E}[\mathbf{C}^{\top}\mathbf{C}]x \ge 0$$
$$\mathbf{E}[x^{T}\mathbf{C}^{\top}\mathbf{C}x] \ge 0$$
$$x^{T}\mathbf{C}^{\top}\mathbf{C}x \ge 0$$
$$(\mathbf{C}x)^{\top}(\mathbf{C}x) \ge 0.$$

[P11]

E47

Define base case:

$$C_{1,2} := C_1 \circ C_2 \in \mathbb{B}^d(\underbrace{(\omega_1 + 1)(\omega_2 + 1) - 1}),$$

$$C_{1,3} := C_1 \circ C_2 \circ C_3 = C_{1,2} \circ C_3,$$

$$\omega_{1,3} := (\omega_{1,2} + 1)(\omega_3 + 1) - 1$$

$$= ((\omega_1 + 1)(\omega_2 + 1) - 1 + 1)(\omega_3 + 1) - 1$$

$$= (\omega_1 + 1)(\omega_2 + 1)(\omega_3 + 1) - 1,$$

$$C_{1,n} := C_1 \circ C_2 \circ \dots \circ C_n = C_{1,n-1} \circ C_n,$$

$$\omega_{1,n} := (\omega_1 + 1)(\omega_2 + 1)\dots(\omega_n + 1) - 1 = (\omega_{1,n-1} + 1)(\omega_n + 1) - 1.$$

By induction, the base case is clear. Next, if n = k, assume

$$C_{1,k} := C_1 \circ C_2 \circ \dots \circ C_k = C_{1,k-1} \circ C_k,$$

$$\omega_{1,k} := (\omega_1 + 1)(\omega_2 + 1)\dots(\omega_k + 1) - 1 = (\omega_{1,k-1} + 1)(\omega_k + 1) - 1$$

is true. Then for n = k + 1:

$$\begin{split} C_{1,k+1} &:= C_1 \circ C_2 \circ \dots \circ C_k \circ C_{k+1} = C_{1,k-1} \circ C_k \circ C_{k+1}, \\ \omega_{1,k+1} &:= (\omega_{1,k}+1)(\omega_{k+1}+1) - 1 = (\omega_1+1)(\omega_2+1)\dots(\omega_k+1)(\omega_{k+1}+1) - 1 \\ &:= ((\omega_{1,k-1}+1)(\omega_k+1) - 1 + 1)(\omega_{k+1}+1) - 1 = (\omega_1+1)(\omega_2+1)\dots(\omega_k+1)(\omega_{k+1}+1) - 1 \\ &:= ((\omega_1+1)(\omega_2+1)\dots(\omega_k+1))(\omega_{k+1}+1) - 1 = (\omega_1+1)(\omega_2+1)\dots(\omega_k+1)(\omega_{k+1}+1) - 1, \\ \omega_{1,k+1} &= (\omega_1+1)(\omega_2+1)\dots(\omega_k+1)(\omega_{k+1}+1) - 1. \end{split}$$

E48

Define

$$\min\{a_i, b_i\} = \begin{cases} a_i, & \text{if } a_i < b_i \\ b_i, & \text{if } a_i > b_i \end{cases}.$$

Thus

$$\sum_{i} \min\{a_i, b_i\} = \begin{cases} \sum_{i} a_i, & \text{if } a_i < b_i, \forall i \\ \sum_{i} b_i, & \text{if } a_i > b_i, \forall i \end{cases}.$$

In case of inequality, define

$$I := \{i | a_i < b_i\},\$$

$$J := \{i | a_i > b_i\}.$$

Thus

$$\sum_{i} \min\{a_i, b_i\} < \begin{cases} \sum_{i} a_i, & \text{if } |I| > |J| \\ \sum_{i} b_i, & \text{if } |J| > |I| \end{cases}.$$