

HW2 (a):

1) Define $z_i \triangleq y_i x_i \Rightarrow f_i(w) = \ln(1 + e^{-w^T z_i})$

$$\nabla f_i(w) = \frac{-e^{-w^T z_i}}{1 + e^{-w^T z_i}} z_i = \left(\frac{1}{1 + e^{-w^T z_i}} - 1 \right) z_i \Rightarrow \|\nabla f_i(w)\|_2 \leq \|z_i\|_2$$

$$\nabla f(w) = \frac{1}{N} \sum_{i=1}^N \nabla f_i(w) + 2\lambda w \Rightarrow \|\nabla f(w)\|_2 \leq \frac{1}{N} \sum_{i=1}^N \|z_i\|_2 + 2\lambda \|w\|_2$$

Hence, if $\|w\|_2 \leq D \Rightarrow \|\nabla f(w)\|_2 \leq B$ for $B = \frac{1}{N} \sum_{i=1}^N \|z_i\|_2 + 2\lambda D$

So, it is Lipschitz.

2) $\nabla^2 f_i(w) = \frac{1}{1 + e^{-w^T z_i}} \left(1 - \frac{1}{1 + e^{-w^T z_i}} \right) z_i z_i^T$

3) $\nabla^2 f(w) = \frac{1}{N} \sum_{i=1}^N \nabla^2 f_i(w) + 2\lambda I$

$$\lambda_{\max}(\nabla^2 f_i(w)) = \frac{1}{1 + e^{-w^T z_i}} \left(1 - \frac{1}{1 + e^{-w^T z_i}} \right) \|z_i\|_2^2 \leq \frac{1}{4} \|z_i\|_2^2$$

$$\lambda_{\min}(\nabla^2 f_i(w)) = 0$$

$$\lambda_{\max}(\nabla^2 f(w)) \leq \frac{1}{N} \sum_{i=1}^N \lambda_{\max}(\nabla^2 f_i(w)) + 2\lambda \leq \frac{1}{4N} \sum_{i=1}^N \|z_i\|_2^2 + 2\lambda$$

$$\lambda_{\min}(\nabla^2 f(w)) \geq \frac{1}{N} \sum_{i=1}^N \lambda_{\min}(\nabla^2 f_i(w)) + 2\lambda = 2\lambda$$

Hence, f is L -smooth with $L = \frac{1}{4N} \sum_{i=1}^N \|z_i\|_2^2 + 2\lambda$

f is μ -strongly convex with $\mu = 2\lambda$

HW2 (b):

$$\text{Var}[g_k(w_k)|w_k] \triangleq \mathbb{E}[\|g_k(w_k)\|_2^2|w_k] - \|\mathbb{E}[g_k(w_k)|w_k]\|_2^2 \leq M + M_V \underbrace{\|\nabla f(w_k)\|_2^2}_{\text{from (3)}}$$

$$\Rightarrow \mathbb{E}[\|g_k(w_k)\|_2^2|w_k] \leq M + M_V \|\nabla f(w_k)\|_2^2 + \|\mathbb{E}[g_k(w_k)|w_k]\|_2^2$$

$$\text{from (2b)} \quad \preceq M + (M_V + c_0^2) \|\nabla f(w_k)\|_2^2$$

Hence, $\boxed{\alpha = M}$, $\boxed{\beta = M_V + c_0^2}$

HW 2(c):

$$W_{k+1} = W_k - \alpha_k g_k(W_k)$$

$$\begin{aligned} f \text{ is } L\text{-smooth} \Rightarrow f(W_{k+1}) &\leq f(W_k) + \nabla f(W_k)^T (W_{k+1} - W_k) + \frac{L}{2} \|W_{k+1} - W_k\|_2^2 \\ &= f(W_k) - \alpha_k \nabla f(W_k)^T g_k(W_k) + \alpha_k^2 \frac{L}{2} \|g_k(W_k)\|_2^2 \end{aligned}$$

Hence,

$$\mathbb{E}[f(W_{k+1}) - f(W_k) | W_k] \leq -\alpha_k \mathbb{E}[\nabla f(W_k)^T g_k(W_k) | W_k] + \alpha_k^2 \frac{L}{2} \mathbb{E}[\|g_k(W_k)\|_2^2 | W_k]$$

$$\begin{aligned} \text{from (2a) and HW2(b)} \quad &\leq -c \alpha_k \|\nabla f(W_k)\|_2^2 + \alpha_k^2 \frac{L}{2} (M + M_G \|\nabla f(W_k)\|_2^2) \\ (M_G \triangleq M + c_0^2) \quad &= -\|\nabla f(W_k)\|_2^2 \alpha_k \left(c - \frac{LM_G}{2} \alpha_k\right) + \alpha_k^2 \frac{LM}{2} \end{aligned}$$

$$\text{if } \forall k \quad \alpha_k \leq \frac{c}{LM_G} \quad \leq -\|\nabla f(W_k)\|_2^2 \alpha_k \frac{c}{2} + \alpha_k^2 \frac{LM}{2}$$

By summing from $k=0$ to $N-1$, we obtain that

$$\begin{aligned} \sum_{k=0}^{N-1} \mathbb{E}[f(W_{k+1}) - f(W_k) | W_k] &\leq -\frac{c}{2} \sum_{k=0}^{N-1} \alpha_k \mathbb{E}[\|\nabla f(W_k)\|_2^2] + \frac{LM}{2} \sum_{k=0}^{N-1} \alpha_k^2 \\ \Rightarrow \mathbb{E}[f(W_N)] - f(W_0) &\leq -\frac{c}{2} \sum_{k=0}^{N-1} \alpha_k \mathbb{E}[\|\nabla f(W_k)\|_2^2] + \frac{LM}{2} \sum_{k=0}^{N-1} \alpha_k^2 \end{aligned}$$

If f has an infimum greater than $-\infty$, the sequence $\mathbb{E}[f(W_N)]$ is always greater than f_{\inf}

$$\underbrace{f_{\inf} - f(W_0)}_{> -\infty} \leq -\frac{c}{2} \sum_{k=0}^{\infty} \alpha_k \mathbb{E}[\|\nabla f(W_k)\|_2^2] + \underbrace{\frac{LM}{2} \sum_{k=0}^{\infty} \alpha_k^2}_{< \infty \text{ because of square summability}}$$

$$\Rightarrow \sum_{k=0}^{\infty} \alpha_k \mathbb{E}[\|\nabla f(W_k)\|_2^2] < \infty \quad \Rightarrow (11) \checkmark$$

If $\sum_{k=0}^{\infty} \alpha_k = \infty$, then

$$\lim_{N \rightarrow \infty} \frac{\sum_{k=0}^N \alpha_k \mathbb{E}[\|\nabla f(W_k)\|_2^2]}{\sum_{k=0}^N \alpha_k} = \frac{\sum_{k=0}^{\infty} \alpha_k \mathbb{E}[\|\nabla f(W_k)\|_2^2]}{\sum_{k=0}^{\infty} \alpha_k} = 0 \Rightarrow (12) \checkmark$$