A APPENDIX

Proof of Lemma 2. Denote W^* as the ground truth diagonal matrix for the training instances, *i.e.*, $W^*_{ii} = 1$ if the i-th instance is a clean instance, otherwise W^*_{ii} . Accordingly, define D_c as the ground truth set of clean instances. For clearness of the presentation, we may drop the subscript t when there is no ambiguation. For mislabeled instances, the output is written in the form of $y_i = r_i + n_i$ where n_i represents the observation noise, and y_i depends on the specific setting we consider. Under this general representation, we can re-write the term w_{t+1} as

$$\begin{split} w_{t+1} &= (\Phi(X)^\top W \Phi(X))^{-1} \Phi(X)^\top W (W^* \Phi(X) w_c^* + (I - W^*) r + e) \\ &= w_c^* + (\Phi(X)^\top W \Phi(X))^{-1} (\Phi(X)^\top W W^* \Phi(X) w_c^* \\ &+ \Phi(X)^\top W r - \Phi(X)^\top W W^* r - \Phi(X)^\top W \Phi(X) w_c^* + \Phi(X) W e) \\ &= w_c^* + (\Phi(X)^\top W \Phi(X))^{-1} \Phi(X)^\top (W W^* - W) (\Phi(X) w_c^* - r - e) \\ &+ (\Phi(X)^\top W \Phi(X))^{-1} \Phi(X)^\top W W^* e \end{split}$$

Therefore, the l_2 distance between the learned parameter and ground truth parameter can be bounded by:

$$\begin{split} \|w_{t+1} - w_c^*\| &= \|(\Phi(X)^\top W \Phi(X))^{-1} \Phi(X)^\top (WW^* - W) \\ &(\Phi(X) w_c^* - r - e) + (\Phi(X)^\top W \Phi(X))^{-1} \Phi(X)^\top WW^* e\|_2 \\ &\leq \underbrace{\|(\Phi(X)^\top W \Phi(X))^{-1}\|_2}_{\nu_1} \cdot \\ &\underbrace{\left(\|\Phi(X)^\top (WW^* - W)(\Phi(X) w_c^* - r - e)\|_2}_{\nu_2} \right. \\ &+ \underbrace{\|\Phi(X)^\top WW^* e\|_2}_{\nu_2} \right) \end{split}$$

where basic spectral norm inequalities and triangle inequalities. For the term v_1 , notice that W selects αN rows of $\Phi(X)$, *i.e.*, $Tr(W) = \alpha N$. Therefore, $v_1 \leq \frac{1}{\Psi^-(\alpha N)}$.

Next, the term v_2 can be bounded as:

$$\begin{split} v_2^2 &= \| (\Phi(X)^\top (W - WW^*) (\Phi(X) w_c^* - r - e) \|_2^2 \\ &= (\Phi(X) w_c^* - r - e)^\top \\ & [(W - WW^*) \Phi(X) \Phi(X)^\top (W - WW^*)] (\Phi(X) w_c^* - r - e) \\ &\leq 2 (\Phi(X) w_c^* - \Phi(X) w_t)^\top \\ & [(W - WW^*) \Phi(X) \Phi(X)^\top (W - WW^*)] (\Phi(X) w_c^* - \Phi(X) w_t) \\ &+ 2 (\Phi(X) w_t - r - e)^\top \\ & [(W - WW^*) \Phi(X) \Phi(X)^\top (W - WW^*)] (\Phi(X) w_t - r - e) \\ &\leq 2 \sigma_{max} (\Phi(X)^\top (W - WW^*) \Phi(X))^2 \| w_c^* - w_t \|_2^2 \\ &+ 2 (\Phi(X) w_t - r - e)^\top \\ & [(W - WW^*) \Phi(X) \Phi(X)^\top (W - WW^*)] (\Phi(X) w_t - r - e) \end{split}$$

For the first term $2\sigma_{max}(\Phi(X)^{\top}(W-WW^*)\Phi(X))^2\|w_c^*-w_t\|_2^2$, let $|D_t\setminus D_c|$ be the number of mislabeled instances in D_t . Then, the eigenvalue is bounded by $\Psi^+(|D_t\setminus D_c|)$. And the last term $2(\Phi(X)w_t-r-e)^{\top}[(W-WW^*)\Phi(X)\Phi(X)^{\top}(W-WW^*)](\Phi(X)w_t-r-e)$ is defined as $\varphi_t := \varphi(D_t,D_c,\|w_c^*-w_t\|_2) = \|\sum_{i\in D\setminus D_c}(\delta(x_i)^{\top}w_t-r_i-e_i)\delta(x_i)\|_2$. The term v_3 can be bounded as:

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$$\begin{aligned} v_3^2 &= \|\Phi(X)^\top W W^* e\|_2 \\ &\leq e^\top \Phi(X) \Phi(X)^\top e \\ &= \sum_{i=1}^d (\sum_{j=1}^N e_j \delta(x_j)_i)^2 \\ &\leq c \sum_{i=1}^N \|\delta(x_i)\|_2^2 \log N \zeta^2 \end{aligned}$$

where the last inequality holds with high probability by the sub-exponential concentration property, and all randomness comes from the measurement noise e. Then, as a summary, combining the results for all three terms, we have:

$$\begin{split} \|w_{t+1} - w_c^*\|_2 & \leq \frac{\sqrt{2}\Psi^+(|D_t \setminus D_c|)}{\Psi^-(\alpha N)} \|w_t - w_c^*\|_2 \\ & + \frac{\sqrt{2}\varphi(D_t, D_c, \|w_c^* - w_t\|_2)}{\Psi^-(\alpha N)} \\ & + c\frac{\sqrt{\sum_{i=1}^N \|(\delta(x_i)\|_2^2 \log N\zeta}}{\Psi^-(\alpha N)} \end{split}$$

The following does the same for the more general non-linear case.

Non-Linear Case. Assume $\pi : \mathbb{R} \to \mathbb{R}$ monotone and differentiable. Assume $\pi'(u) \in [a, b]$ for all $u \in \mathbb{R}$, where a, b are positive constants. Define $F : \mathbb{R}^n \to \mathbb{R}^n$ as an entry-wise f(.)-operation. Denote learning rate as η .

$$\begin{split} w_{t+1} &= w_t - \frac{\eta}{\alpha N} \sum_{i \in D_t} (f(\delta(x_i)^\top w_t) - y_i) \cdot f^{'}(\delta(x_i)^\top w_t) \cdot \delta(x_i) \\ &= w_t - \frac{\eta}{\alpha N} \Phi(X)^\top \mathrm{Diag}(F^{'}(\Phi(X)w_t)) W_t(F(\Phi(X)w_t) - y) \\ &= w_t - \frac{\eta}{\alpha N} \Phi(X)^\top \mathrm{Diag}(F^{'}(\Phi(X)w_t)) W_t(F(\Phi(X)w_t) - W^*F(\Phi(X)w_c^*) - (I - W^*)(r + e) - W^*e) \\ &= w_t - \frac{\eta}{\alpha N} \Phi(X)^\top \mathrm{Diag}(F^{'}(\Phi(X)w_t)) W_t(F(\Phi(X)w_t) - W^*F(\Phi(X)w_c^*) - (I - W^*)F(\Phi(X)w_c^*)) \\ &- \frac{\eta}{\alpha N} \Phi(X)^\top \mathrm{Diag}(F^{'}(\Phi(X)w_t)) W_t((I - W^*)F(\Phi(X)w_c^*) - (I - W^*)(r + e) - W^*e) \\ &= w_t - \frac{\eta}{\alpha N} \Phi(X)^\top \mathrm{Diag}(F^{'}(\Phi(X)w_t)) W_t(F(\Phi(X)w_t) - F(\Phi(X)w_c^*)) \\ &- \frac{\eta}{\alpha N} \Phi(X)^\top \mathrm{Diag}(F^{'}(\Phi(X)w_t)) (W_t - W_t W^*)(F(\Phi(X)w_c^*) - r - e) \\ &+ \frac{\eta}{\alpha N} \Phi(X)^\top \mathrm{Diag}(F^{'}(\Phi(X)w_t)) W_t W^*e \end{split}$$

We simplify the notation using $H_t \triangleq \text{Diag}(F^{'}(\Phi(X)w_t))$. Also, by mean value theorem, for any a,b, there exists some $c \in [a,b]$, such that $\frac{f(b)-f(a)}{b-a} = f^{'}(c)$. Therefore, for the term $F(\Phi(X)w_t) - F(\Phi(X)w_c^*)$, there exists a diagonal matrix C_t , such that $F(\Phi(X)w_t) - F(\Phi(X)w_c^*) = C_t\Phi(X)(w_t - w_c^*)$. Therefore, we have

$$\begin{aligned} \|w_{t+1} - w_c^*\|_2 &\leq \left(\underbrace{1 - \frac{\eta}{\alpha N} \Phi(X)^\top H_t W_t C_t \Phi(X)}_{U_1}\right) \|w_t - w_c^*\|_2 \\ &+ \frac{\eta}{\alpha N} \underbrace{\|\Phi(X)^\top H_t (W_t - W_t W^*) (F(\Phi(X) w_c^*) - r - e)\|_2}_{U_2} \\ &+ \frac{\eta}{\alpha N} \underbrace{\|\Phi(X) H_t W_t W^* e\|_2}_{U_3} \end{aligned}$$

Here,

$$U_1 \le 1 - \eta a^2 \frac{\Psi^-(\alpha N)}{\alpha N}, U_3 \le b \xi_t \zeta$$

For U_2 , define $\hat{\delta_t}$ similar to δ_t :

$$\hat{\varphi_t} = \| \sum_{i \in D_t \setminus D_c} (\pi(\delta(x_i)^\top w_c^*) - r_i - e_i) \pi'(\delta(x_i)^\top w_c^*) \delta(x_i) \| \cdot$$

As a result, we have:

$$\|w_{t+1} - w_c^*\|_2 \le \left(1 - \frac{\eta}{\alpha N} a^2 \Psi^-(\alpha N)\right) \|w_t - w_c^*\|_2 + \eta \frac{\hat{\varphi}_t + \xi_t b \zeta}{\alpha N}$$