

## A APPENDICES

### A.1 PROOF OF THEOREM 2

*Proof.* Let's try expanding  $\frac{\partial^2 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \epsilon^2}$  for the first time:

$$\begin{aligned}
\frac{\partial^2 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \epsilon^2} &= \frac{\partial \left( \frac{\partial L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \epsilon} \right)}{\partial \epsilon} \\
&= \frac{\partial \left( \left( \frac{\partial L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta} \right)^T \cdot \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} \right)}{\partial \epsilon} \\
&= \left( \frac{\partial^2 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta \partial \epsilon} \right)^T \cdot \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} + \left( \frac{\partial L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta} \right)^T \cdot \frac{\partial^2 \hat{\theta}_\epsilon}{\partial \epsilon^2} \\
&= \left( \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} \right)^T \cdot \frac{\partial^2 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta^2} \cdot \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} + \left( \frac{\partial L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta} \right)^T \cdot \frac{\partial^2 \hat{\theta}_\epsilon}{\partial \epsilon^2}
\end{aligned} \tag{20}$$

To go on, we need to expand  $\frac{\partial^2 \hat{\theta}_\epsilon}{\partial \epsilon^2}$ , which is as follows:

$$\begin{aligned}
\frac{\partial^2 \hat{\theta}_\epsilon}{\partial \epsilon^2} &= \frac{\partial \left( \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} \right)}{\partial \epsilon} \\
&= \frac{\partial \left( -W_\epsilon^{-1} \cdot \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right) \right)}{\partial \epsilon} \\
&= -\frac{\partial (W_\epsilon^{-1})}{\partial \epsilon} \cdot \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right) - W_\epsilon^{-1} \cdot \frac{\partial \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right)}{\partial \epsilon} \\
&= W_\epsilon^{-1} \cdot \frac{\partial W_\epsilon}{\partial \epsilon} \cdot W_\epsilon^{-1} \cdot \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right) - W_\epsilon^{-1} \cdot \left( \frac{1}{n} \cdot \frac{\partial \mathbf{b}'_\epsilon}{\partial \epsilon} + \frac{\delta'_\epsilon}{\partial \epsilon} \cdot \hat{\theta}_\epsilon + \delta'_\epsilon \cdot \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} \right) \\
&= W_\epsilon^{-1} \cdot \left( \frac{\partial^3 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta^3} + \delta'_\epsilon \cdot \mathbf{I} \right) \cdot \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} - W_\epsilon^{-1} \cdot \left( \frac{1}{n} \cdot \frac{\partial \mathbf{b}'_\epsilon}{\partial \epsilon} + \frac{\delta'_\epsilon}{\partial \epsilon} \cdot \hat{\theta}_\epsilon + \delta'_\epsilon \cdot \frac{\partial \hat{\theta}_\epsilon}{\partial \epsilon} \right)
\end{aligned} \tag{21}$$

Finally, let  $\mathbf{b}''_\epsilon = \frac{\partial \mathbf{b}'_\epsilon}{\partial \epsilon}$ ,  $\delta''_\epsilon = \frac{\partial \delta'_\epsilon}{\partial \epsilon}$ ,  $H_{\hat{\theta}_\epsilon} = \frac{\partial^2 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta^2}$  and  $T_{\hat{\theta}_\epsilon} = \frac{\partial^3 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \theta^3}$

$$\begin{aligned}
\frac{\partial^2 L(\hat{\theta}_\epsilon, \mathcal{D})}{\partial \epsilon^2} &= \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right)^T \cdot W_\epsilon^{-1} \cdot H_{\hat{\theta}_\epsilon} \cdot W_\epsilon^{-1} \cdot \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right) \\
&\quad - (\nabla_\theta L(\hat{\theta}_\epsilon, \mathcal{D}))^T \cdot W_\epsilon^{-1} \cdot (T_{\hat{\theta}_\epsilon} + \delta'_\epsilon \cdot \mathbf{I}) \cdot W_\epsilon^{-1} \cdot \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right) \\
&\quad - (\nabla_\theta L(\hat{\theta}_\epsilon, \mathcal{D}))^T \cdot W_\epsilon^{-1} \cdot \left( \frac{1}{n} \cdot \mathbf{b}''_\epsilon + \delta''_\epsilon \cdot \hat{\theta}_\epsilon \right) \\
&\quad + (\nabla_\theta L(\hat{\theta}_\epsilon, \mathcal{D}))^T \cdot W_\epsilon^{-1} \cdot \delta'_\epsilon \cdot W_\epsilon^{-1} \cdot \left( \frac{1}{n} \mathbf{b}'_\epsilon + \delta'_\epsilon \cdot \hat{\theta}_\epsilon \right)
\end{aligned} \tag{22}$$

which concludes the proof.  $\square$

### A.2 PROOF OF THEOREM 3

*Proof.* According to the assumptions in Theorem 3, the error bound depends on  $\mathbf{b}'_\epsilon$ ,  $\mathbf{b}''_\epsilon$ ,  $\delta_{\hat{\theta}_\epsilon}$ ,  $\delta'_{\hat{\theta}_\epsilon}$  and  $\delta''_{\hat{\theta}_\epsilon}$ .

We first bound  $\mathbf{b}'_\epsilon$  and  $\mathbf{b}''_\epsilon$ . Note that  $\beta_\epsilon$  is defined as 4 and  $\mathbf{b}'_\epsilon$  can be calculated by 13, it is easy to figure out the following:  $\|\mathbf{b}'_\epsilon\| = \mathcal{O}(\frac{1}{\epsilon^2})$ ,  $\|\mathbf{b}''_\epsilon\| = \mathcal{O}(\frac{1}{\epsilon^3})$ .

Next, we will bound  $\delta_{\hat{\theta}_\epsilon}$ ,  $\delta'_{\hat{\theta}_\epsilon}$  and  $\delta''_{\hat{\theta}_\epsilon}$ . With the fact that  $\epsilon < 1$ , we have  $e^{\frac{\epsilon}{4}} - 1 = \frac{\epsilon}{4} + o(\epsilon)$ . With the definition of  $\delta_{\hat{\theta}_\epsilon}$  in 4, we have  $\|\delta_{\hat{\theta}_\epsilon}\| = \mathcal{O}(\frac{1}{n\epsilon})$ ,  $\|\delta'_{\hat{\theta}_\epsilon}\| = \mathcal{O}(\frac{1}{n\epsilon^2})$  and  $\|\delta''_{\hat{\theta}_\epsilon}\| = \mathcal{O}(\frac{1}{n\epsilon^3})$ .

Finally, we get our error bound as follows:

$$\mathcal{O} \left( \frac{(\epsilon' - \epsilon)^2}{n^2 \epsilon^4} \right),$$

which concludes the proof.  $\square$

## A.3 SUPPLEMENTARY EXPERIMENTAL RESULTS

Table 2: Average error of estimated loss for measuring point from  $\epsilon = 0.05$  to  $\epsilon = 0.35$ .

$\epsilon$	0.05	0.10	0.15	0.20	0.25	0.30	0.35
Adlut-LR	0.01982	0.00823	0.00908	0.00408	<b>0.00003</b>	0.00033	0.00004
Adlut-SVM	0.08188	0.02052	0.00327	0.00315	0.00445	0.00367	0.00385
Kddcup-LR	0.05884	0.02393	0.00930	0.00721	0.00763	0.00845	0.00960
Kddcup-SVM	0.04735	0.00866	0.00798	0.00793	0.00851	0.01052	0.01143
Gisette-LR	0.18427	0.01482	0.01181	0.02125	0.01931	<b>0.00302</b>	0.00810
Gisette-SVM	0.16245	0.00571	0.02480	0.00822	0.00567	0.01023	<b>0.00550</b>

Table 3: Average error of estimated loss for measuring point from  $\epsilon = 0.40$  to  $\epsilon = 0.70$ .

$\epsilon$	0.40	0.45	0.50	0.55	0.60	0.65	0.70
Adlut-LR	0.00037	0.00293	0.00309	0.00283	0.00339	0.00623	0.00643
Adlut-SVM	0.00558	<b>0.00298</b>	0.00585	0.00890	0.00833	0.00711	0.00546
Kddcup-LR	0.00677	<b>0.00283</b>	0.00778	0.00442	0.00768	0.01010	0.01274
Kddcup-SVM	0.00954	0.00848	0.00967	0.00486	0.00363	0.00154	0.00200
Gisette-LR	0.00372	0.00394	0.01361	0.01420	0.01423	0.02461	0.03560
Gisette-SVM	0.00712	0.00823	0.00823	0.00961	0.00926	0.01188	0.01327

Table 4: Average error of estimated loss for measuring point from  $\epsilon = 0.75$  to  $\epsilon = 1$ .

$\epsilon$	0.75	0.80	0.85	0.90	0.95	1.00
Adlut-LR	0.00649	0.01169	0.01480	0.01485	0.01675	0.02276
Adlut-SVM	0.00770	0.00780	0.00840	0.01142	0.01053	0.01020
Kddcup-LR	0.01186	0.01288	0.01200	0.01149	0.01147	0.01213
Kddcup-SVM	0.00112	0.00137	<b>0.00038</b>	0.00134	0.00192	0.00250
Gisette-LR	0.04637	0.04476	0.03693	0.03683	0.03308	0.03828
Gisette-SVM	0.01745	0.03641	0.02911	0.03548	0.04540	0.03603

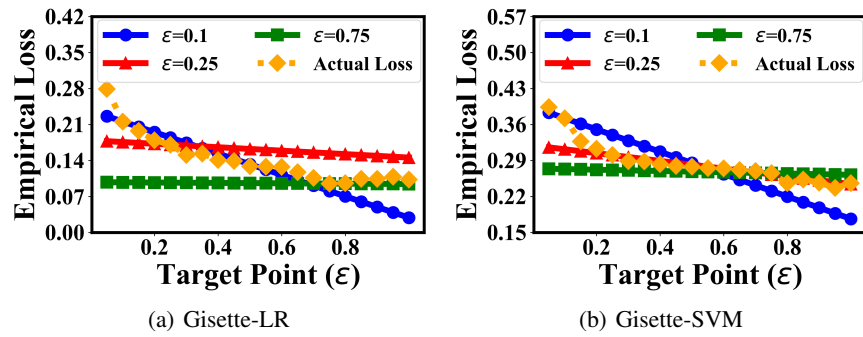


Figure 3: Performance of our approximation approach on Gisette with logistic regression (LR) loss and Huber SVM (SVM) loss