

22.4.2020 ERM for student-t likelihood

$$\operatorname{argmax}_w \hat{P}(y_{1:n} | x_{1:n}, w)$$

$$= \operatorname{argmin}_w - \sum_{i=1}^n \log \hat{P}(y_i | x_i, w)$$

$$= \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\sqrt{\pi} \nu \sigma^2 \Gamma\left(\frac{\nu}{2}\right)} \left(1 + \frac{(y_i - w^T x_i)^2}{\nu \sigma^2}\right)^{-\frac{\nu+1}{2}}$$

Normalization constant
 $Z^{-1}(\nu, \sigma^2)$

$$= \operatorname{argmin}_w - \sum_{i=1}^n \log Z^{-1}(\nu, \sigma^2) \left(1 + \frac{(y_i - w^T x_i)^2}{\nu \sigma^2}\right)^{-\frac{\nu+1}{2}}$$

$$= \operatorname{argmin}_w - n \cdot \log Z^{-1}(\nu, \sigma^2) + \sum_{i=1}^n \left(\frac{\nu+1}{2}\right) \log \left(1 + \frac{(y_i - w^T x_i)^2}{\nu \sigma^2}\right)$$

$$= \operatorname{argmin}_w \sum_{i=1}^n \log \left(1 + \frac{(y_i - w^T x_i)^2}{\nu \sigma^2}\right)$$

$$= \operatorname{argmin}_w \sum_{i=1}^n \log \left(1 + \frac{(y_i - w^T x_i)^2}{c}\right) \quad \text{for } c = \nu \cdot \sigma^2$$

$$\ell_{\text{student}}(w; y_i, x_i) \leq \frac{(y_i - w^T x_i)^2}{c} = \frac{1}{c} \cdot \ell_2(w; y_i, x_i)$$

Since $\log(1+x) \leq x$

$\log(1+x) \approx x$ for $x \approx 0$; $\log(1+x) \ll x$ for $x \gg 0$
 \Rightarrow Robustness against outliers