Machine Learning: Project 1 (Addendum)

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1 Logistic regression for labels in $\{-1,1\}$

We start from the standard negative log-likelihood from [1] for labels $y_n \in \{0, 1\}$:

$$\mathcal{L}(\boldsymbol{w}) = \sum_{n=1}^{N} \ln \left[1 + \exp(\boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}) \right] - y_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}$$

$$= \sum_{n=1}^{N} \ln \left[\frac{1 + \exp(\boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w})}{\exp(y_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w})} \right]$$

$$= \sum_{n=1}^{N} \ln \left[\exp(-\underbrace{y_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}}_{n}) \right]$$

$$= \exp(\underbrace{(1 - y_{n}) \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}}_{n}) \right]$$

$$= \sum_{n=1}^{N} \ln \left[1 + \exp(-\hat{y}_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}) \right]$$

$$= \sum_{n=1}^{N} \ln \left[1 + \exp(-\hat{y}_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}) \right]$$
(1)

where in the last step we replaced y_n with $\hat{y}_n \in \{-1,1\}$. The gradient of this expression is

$$\nabla \mathcal{L}(\boldsymbol{w}) = \sum_{n=1}^{N} \frac{\exp\left(-\hat{\mathbf{y}}_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}\right)}{1 + \exp\left(-\hat{\mathbf{y}}_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}\right)} (-y_{n} \boldsymbol{x}_{n})$$

$$= \sum_{n=1}^{N} \frac{-y_{n} \boldsymbol{x}_{n}}{1 + \exp\left(\hat{\mathbf{y}}_{n} \boldsymbol{x}_{n}^{\mathsf{T}} \boldsymbol{w}\right)}$$

$$= -\boldsymbol{X}^{\mathsf{T}} \left[\boldsymbol{y} \odot \frac{1}{1 + \exp\left(\hat{\boldsymbol{y}} \odot (\boldsymbol{X} \boldsymbol{w})\right)} \right] \quad (2)$$

Here, ⊙ denotes element-wise multiplication (the Hadamard product).

References

[1] *Jaggi M., Urbanke R., Khan M. E.* Machine Learning (CS-433): Lecture notes. 2021. https://github.com/epfml/ML_course.