
Dropout Distillation Supplementary Material

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Abstract

This document provides the proof of Theorem 1 in the main paper (Rota Bulò et al., 2016).

Proof of Thm.1. The proof of *i)* follows by simple application of Jensen’s inequality given the convexity assumption of ℓ :

$$\begin{aligned} J(q) &= \mathbb{E}_{\mathbf{x}} [\ell(f_{\text{dropout}}(\mathbf{x}), q(\mathbf{x}))] \\ &\leq \mathbb{E}_{\mathbf{x}, \sigma} [\ell(f_{\Theta^*, \sigma}(\mathbf{x}), q(\mathbf{x}))] = J'(q). \end{aligned}$$

As for *ii)*, by exploiting the expression for ℓ in terms of g_1, g_2, g_3 we have

$$\begin{aligned} J(q) &= \mathbb{E}_{\mathbf{x}} [g_1(f_{\text{dropout}}(\mathbf{x})) + g_2(q(\mathbf{x})) \\ &\quad + f_{\text{dropout}}(\mathbf{x})^\top g_3(q(\mathbf{x}))] \\ &= \mathbb{E}_{\mathbf{x}, \sigma} [g_1(f_{\Theta^*, \sigma}(\mathbf{x})) + g_2(q(\mathbf{x})) \\ &\quad + f_{\Theta^*, \sigma}(\mathbf{x})^\top g_3(q(\mathbf{x}))] \\ &\quad - \underbrace{\mathbb{E}_{\mathbf{x}, \sigma} [g_1(f_{\Theta^*, \sigma}(\mathbf{x}))]}_{\Delta} + \mathbb{E}_{\mathbf{x}} [g_1(f_{\text{dropout}}(\mathbf{x}))] \\ &= \mathbb{E}_{\mathbf{x}, \sigma} [\ell(f_{\Theta^*, \sigma}(\mathbf{x}), q(\mathbf{x}))] + \Delta \\ &= J'(q) + \Delta, \end{aligned}$$

where Δ is a constant that does not depend on q .

The same derivations hold true if we replace $\mathbb{E}_{\mathbf{x}} [\cdot]$ with an empirical average. \square

References

Rota Bulò, S., Porzi, L., and Kotschieder, P. Dropout distillation. In *Int. Conf. on Machine Learning*, 2016.

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