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Comparative Analysis of Pre-Training Models for Low-Resource Languages

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Abstract: Unlike major languages low-resource languages can suffer from a lack of expert-annotated benchmark and corpora resources, which makes it difficult to correctly apply state-of-the-art techniques. In this study, we take two approaches to addressing this shortage problem for the low-resource Filipino language. First, we provide a brand-new benchmark language modeling dataset in Filipino, WikiText-TL-39. Second, we show that language model finetuning methods such as BERT and ULMFiT can be used to consistently train robust classifiers in low-resource environments, with a validation error increase of no more than 0.0782 when finetuning with a privately owned sentiment dataset and reducing the number of training examples from 10K to 1K.

Keywords: Language Model, Finetuning Techniques, Evaluation, Low-Resource Languages.

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