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Multimodal Inverse Cloze Task for Knowledge-based Visual Question Answering

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We present a new pre-training method, Multimodal Inverse Cloze Task, for Knowledge-based Visual Question Answering about named Entities (KVQAE, [2]). KVQAE is a recently introduced task that consists, as illustrated by Figure 1 in answering questions about named entities grounded in a visual context using a textual Knowledge Base. Therefore, the interaction between the modalities in this task is paramount and must be captured with complex fusion models. As these models require a lot of training data, we design this pre-training task from existing work in textual Question Answering [1]. It consists in considering a sentence as a pseudo-question and its context as a pseudo-relevant passage and is extended by considering images near texts in multimodal documents. Our method is applicable to different neural network architectures and leads to a 9% relative-MRR and 15% relative-F1 gain for retrieval and reading comprehension, respectively, over a no-pre-training baseline.

Visual Question (input) Relevant visual passage in the Knowledge Base "Which constituency did this man represent when he was Prime Minister?" Relevant visual passage in the Knowledge Base "Macmillan indeed lost Stockton in the landslide Labour victory of 1945, but returned to Parliament in the November 1945 by-election in Bromley."

Fig. 1: Example of a question about a named entity with a visual context along with relevant passages from a Knowledge Base.

Keywords: Visual Question Answering \cdot Named Entities \cdot Pre-training \cdot Multimodal Fusion.

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