





# Multimodal Inverse Cloze Task for Knowledge-based Visual Question Answering (KVQAE)

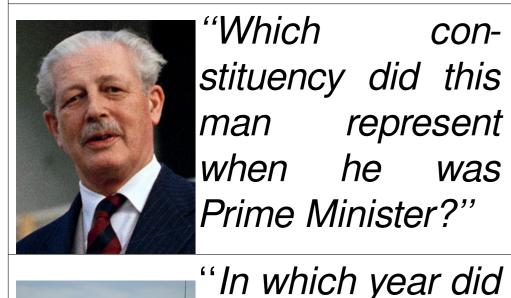
Paul Lerner<sup>1</sup>, Olivier Ferret<sup>2</sup>, Camille Guinaudeau<sup>1</sup>

 $^1$ Université Paris-Saclay, CNRS, LISN, 91400, Orsay, France,  $^2$ Université Paris-Saclay, CEA, List, F-91120, Palaiseau, France Contact: lerner@lisn.fr

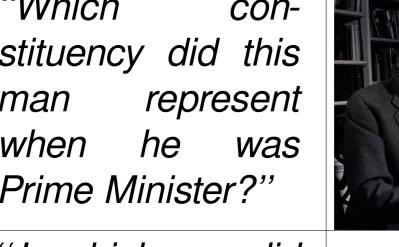
## KVQAE and ViQuAE Dataset and Knowledge Base

KVQAE = Answer questions about named entities grounded in a visual context using a Knowledge Base (KB).

# Visual Question (input) Relevant visual passage in the KB

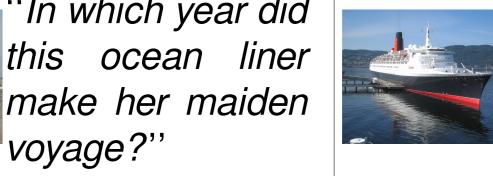


voyage?"





con- "Macmillan indeed lost Stockton in the landslide Labour victory of 1945, but returned to Parliament in the November 1945 by-election in **Brom-**

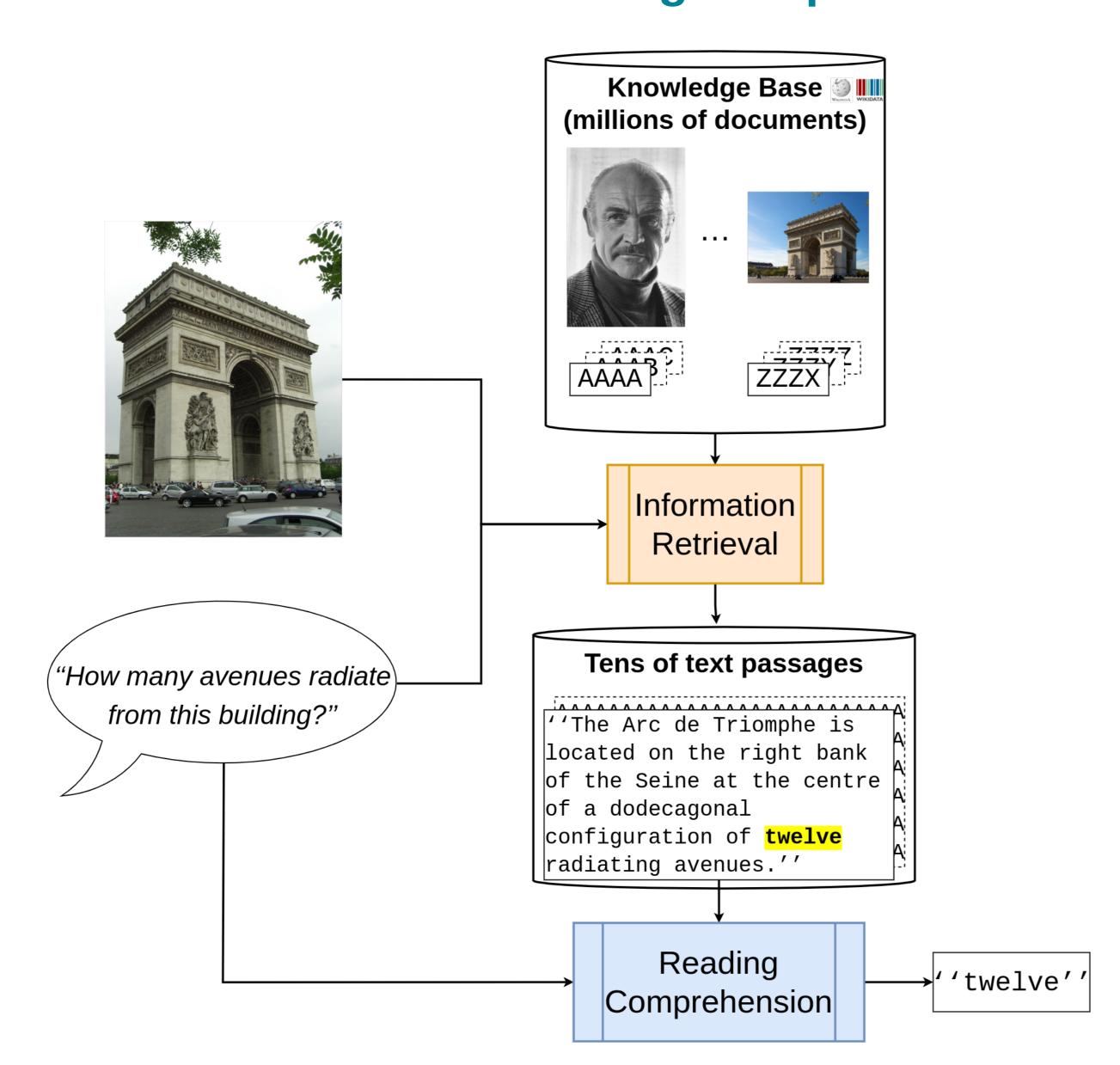


"Queen Elizabeth 2, often referred to simply as QE2, is a floating hotel and retired ocean liner built for the Cunard Line which was operated by Cunard as both a transatlantic liner and a cruise ship from **1969** to 2008."

ViQuAE: dataset of 3.7K visual questions (Lerner et al., 2022). Knowledge Base:

- 1.5M Wikipedia articles with images
- Split in 12M text passages of 100 words

# Information Retrieval + Reading Comprehension



Our focus: Information Retrieval (efficiently retrieves relevant information from the KB). Requires to combine text and image.

Late fusion baseline: results of text and image search fused at the score-level through linear interpolation. Neglects interaction between the modalities.

- Text search: DPR (Karpukhin et al., 2020): built upon two BERT models (Devlin et al., 2019)
- Image search: ResNet-50 (He et al., 2016) trained in the CLIP framework (Radford et al., 2021)

#### Acknowledgements

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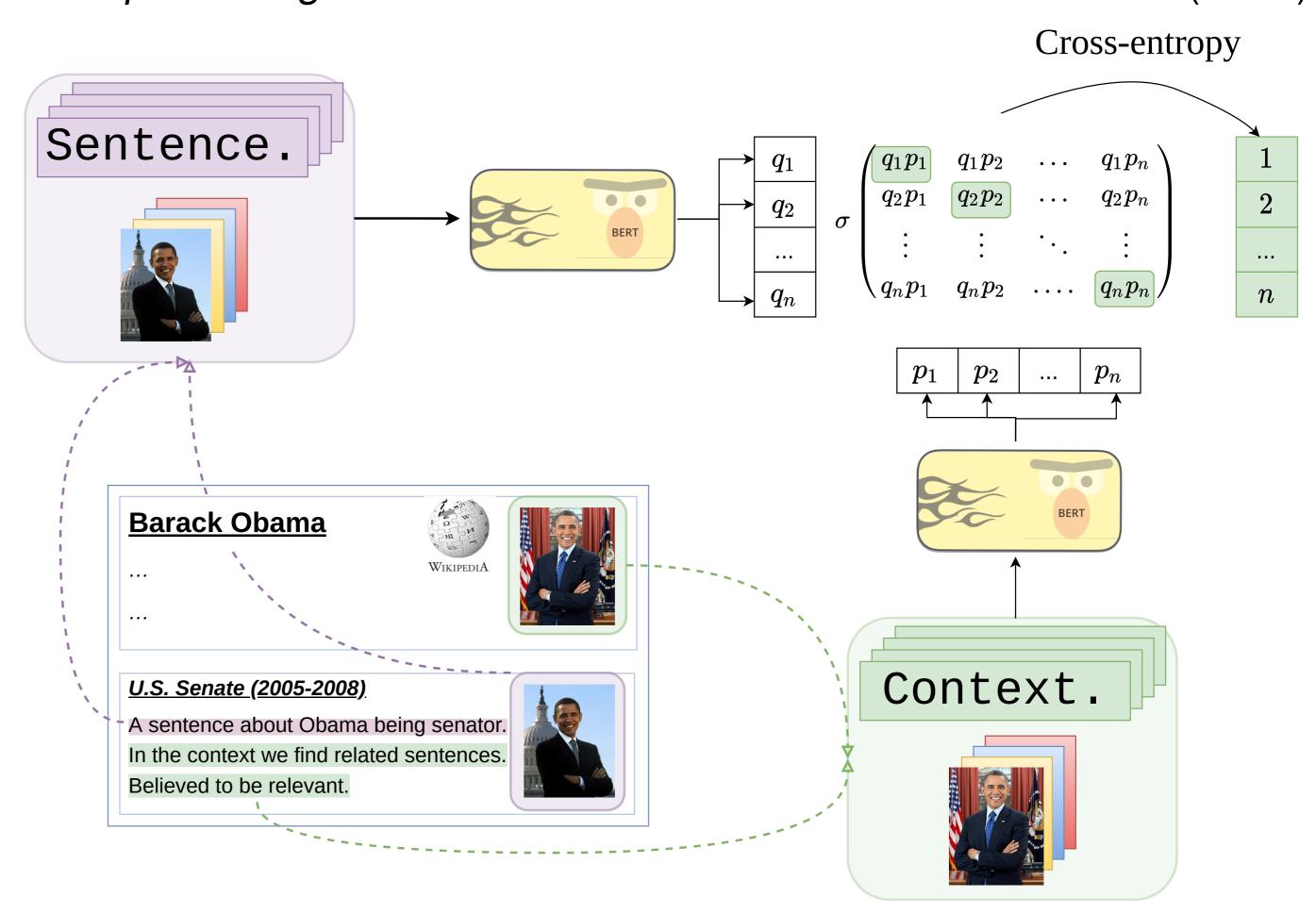
## References

Paul Lerner, Olivier Ferret, Camille Guinaudeau, Hervé Le Borgne, Romaric Besançon, Jose G Moreno, and Jesús Lovón Melgarejo. 2022. Vi-QuAE, a dataset for knowledge-based visual question answering about named entities. In Proceedings of The 45th International ACM SIGIR Conference on Research and Development in Information Retrieval, SIGIR'22, New York, NY, USA. Association for Computing Machinery.

## Multimodal Inverse Cloze Task (ICT)

ViQuAE: small dataset, **early fusion**: complex models

⇒ pretraining with Multimodal ICT. Extended from Lee et al. (2019).



#### **Quantitative Results**

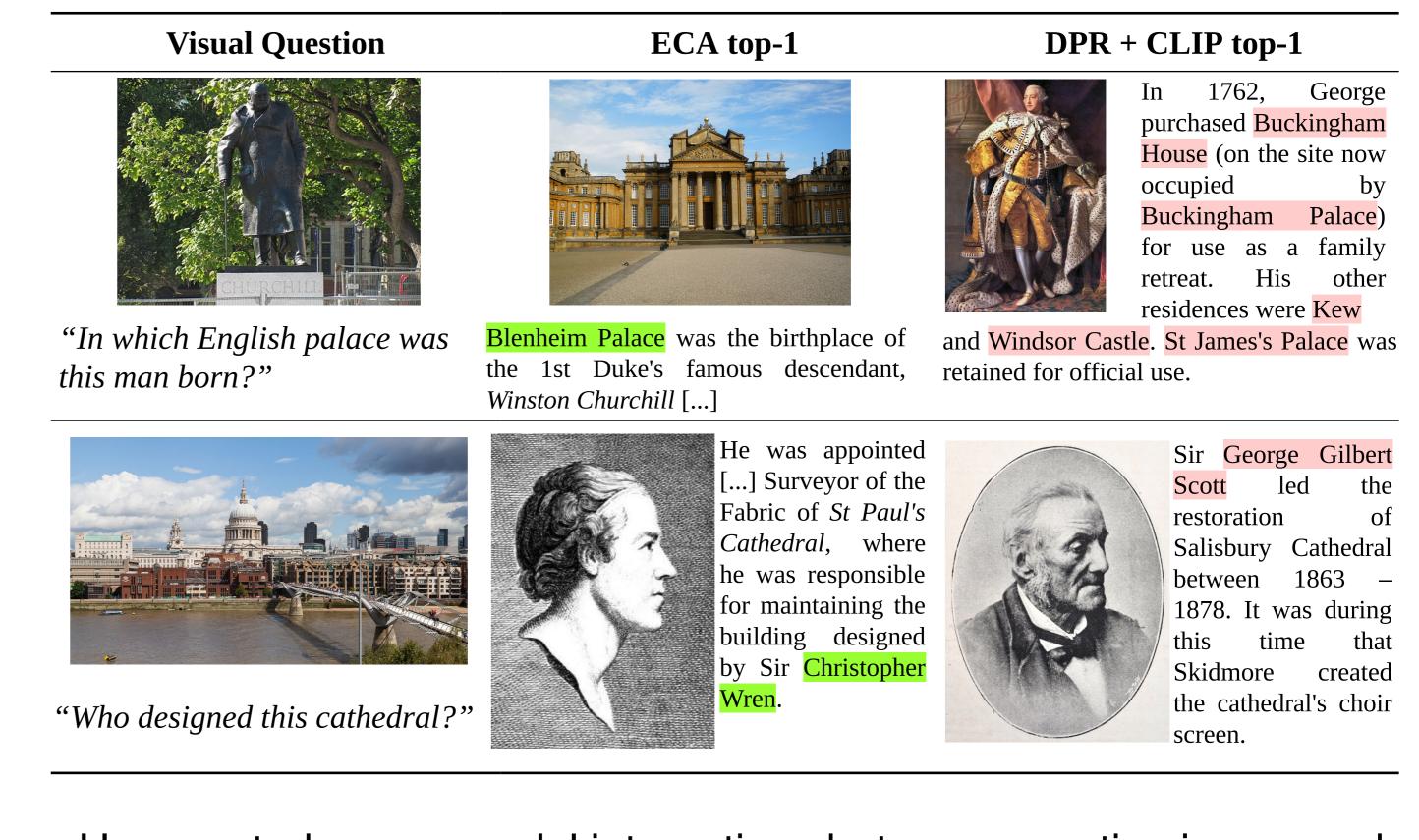
Two metrics for two evaluations:

- Mean Reciprocal Rank (MRR) for Information Retrieval (IR): directly evaluates the relevance of the retrieved visual passages
- F1-score (bag-of-words level) for Reading Comprehension (RC): evaluates the answer extracted by the reader module of Lerner et al.

Model	MRR (IR)	F1 (RC)
DPR (text-only)	32.8	20.1
DPR + CLIP (late fusion)	34.5	22.3
Early Cross-Attention (ECA)	37.8	24.4
Intermediate Linear Fusion (ILF)	37.3	25.4

- Both models outperform late fusion, thanks to cross-modal interactions
- Unexpected: ILF performs on-par with ECA

#### **Qualitative Results**



- Unexpected: cross-modal interactions between question image and passage text
- Interactions between question text and image: harder to assess, not suggested by quantitative results (ILF  $\approx$  ECA)