





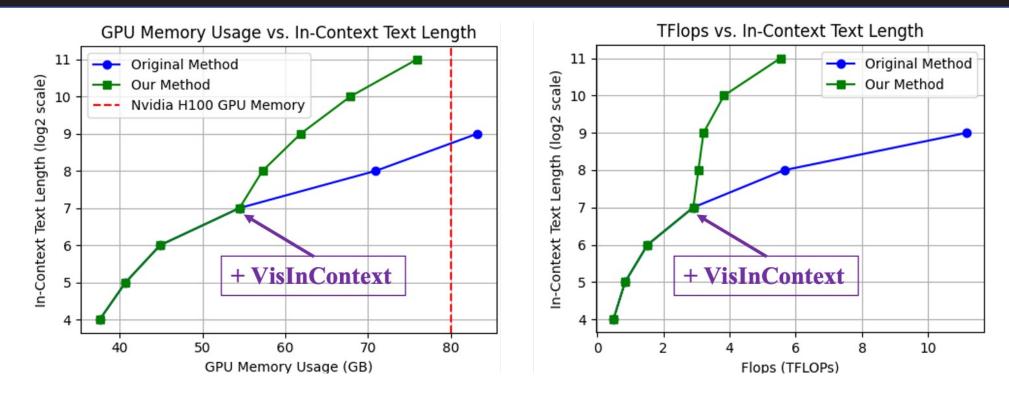




# Leveraging Visual Tokens for Extended Text Contexts in Multi – Modal Learning

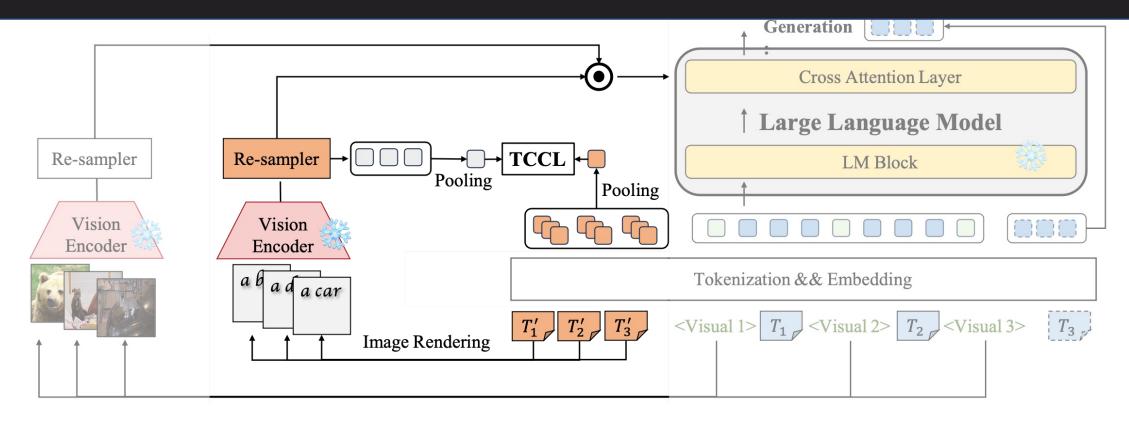
Presented by [Alex] Jinpeng Wang 11  $11^{th}$ , 2024

#### Motivation



- Existing MLLMs usually exploit a much lighter visual encoders, compared to its text decoders
- Visual encoders trained on paired image-text data also exhibit emergent OCR capabilities.
- Convert long textual content into images and use the visual encoders to extract textual representations.

#### Method



- We convert these omitted text context into visual signals by rendering them into image
- VisInContext pipeline builds upon the Flamingo model for in-context few-shot modeling (represented in gray)
- VisInContext processes interleaved image-text data by rendering portions of the in-context text into images
- Maintains the *Text Token Length* of the model while allowing for a significantly extended *In-context Text Length*.

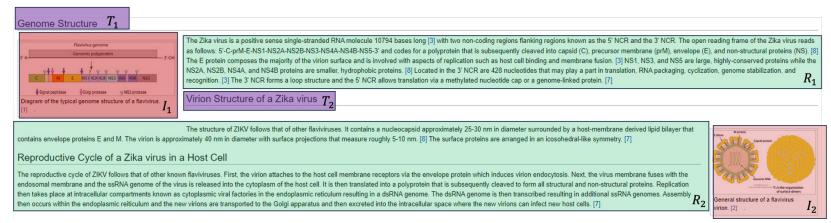


(a). Rendered Text

(b). Token Prediction Accuracy on Validation Set

VisInContext significantly improves the OCR ability of LLM.

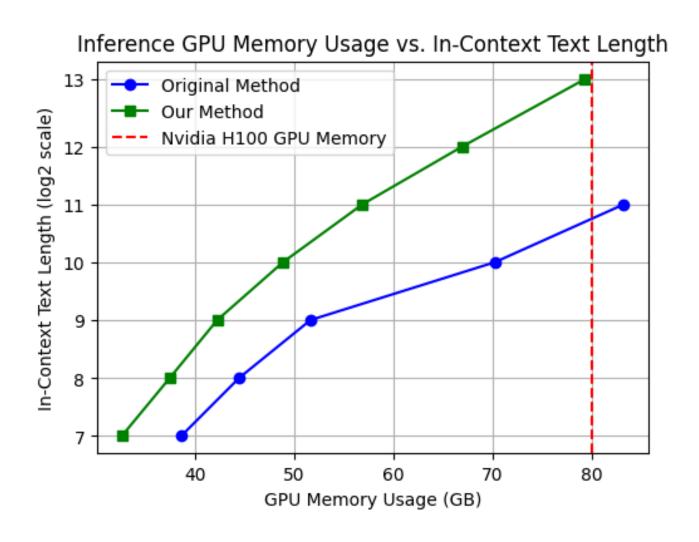
VisInContext achieves significantly better results in predicting words in visual images, even when the fonts are difficult to recognize.

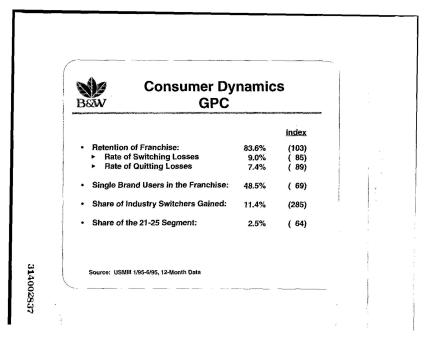


Sequential multi-modal retrieval example.

Visual Input	Text Input	<b>Surrounding Text Input</b>	Seq-I	Seq-T
Raw Image	Raw Text	-	16.3	64.8
Raw Image	Raw Text	Raw Text	18.9	67.5
Raw Image	Raw Text	Rendered Text Image	22.7	66.5

The model pretrain with VisInContext significantly improves sequence understanding ability.





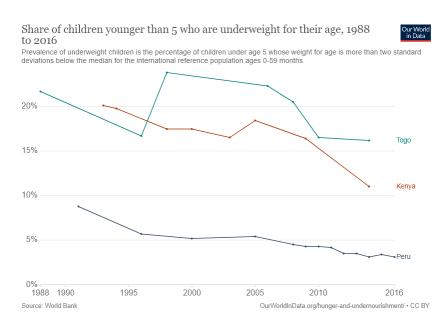
Source: https://www.industrydocuments.ucsf.edu/docs/rzbj0037

**Q:** What is the percentage of the share of

the 21-25 segment?

**Baseline:** 64

+VisInContext: 2.5%



**Q:** Which country is represented by the red line?

**Baseline:** The red line is Peru

+VisInContext : Kenya

#### Contribution

- 1. We introduce Visualized In- Context Text Processing (VisInContext), a novel method that increases in-context text length using visual tokens.
- 2. We demonstrate that VisInContext is **effective for both training and inference stage** with much lower computational cost.
- 3. With extended text context brought by VisInContext, our model improves the average **in-context few-shot performance** from 55.8% to 57.8% over the competing baseline.
- 4. As a byproduct, our method also shows great potential in document understanding on popular document QA tasks and our newly proposed sequential document retrieval task.