





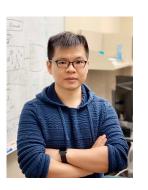
Modality-Specialized Synergizers for Interleaved Vision-Language Generalists

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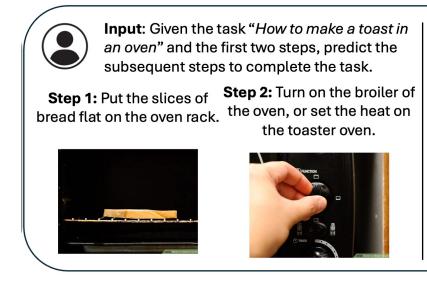






Interleaved Text-and-Image Generation

Interleaved generation^[1,2] that requires models generating **both text and images** is an increasingly important task in multimodal learning.





Output: Here're the subsequent steps:

Step 3: Use tongs to flip the Step 4: Remove the bread over half-way through toast from the oven the toasting.





[2] Emu3: Next-Token Prediction is All You Need. Emu3 Team at BAAI. 2024.

^[1] Chameleon: Mixed-Modal Early-Fusion Foundation Models. Chameleon Team at Meta. 2024.



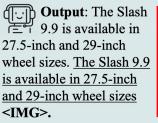




Challenges in Interleaved Text-and-Image Generation

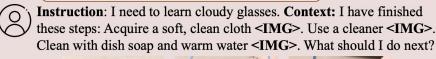
Existing models use the same architecture with the same set of parameters to process both text and images, which may be insufficient to model the distinct inductive biases in each modality given their intrinsic discrepancy.







Weak Instruction-Following Capability













Modality-Specialized Synergizers (MoSS)

Introduce modality-specialized modules to model different inductive biases.

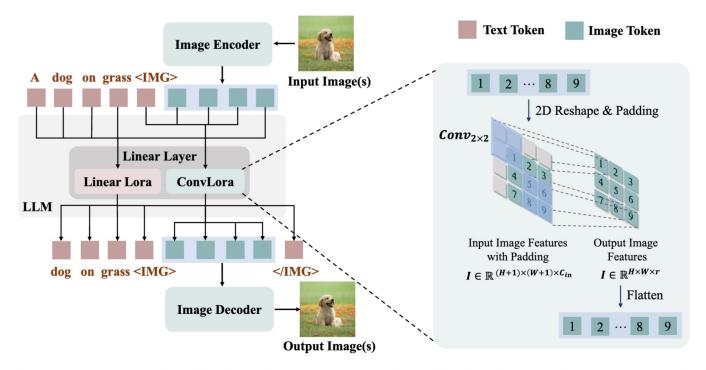


Figure 2: An autoregressive VLG with our proposed MoSS added to its linear layers. The linear LoRA on the left side is specialized to generate text tokens and the Convolutional LoRA on the right side is specialized to generate image patches. On the right handside, we show the details of convolutional operation applied to autoregressively generate image tokens. Best viewed in color.





• We also LeafInstruct, first open-sourced interleaved instruction tuning dataset comprising 184,982 high-quality instances on more than 10 diverse domains.

Instruction: Here are 5 images on the following content: {caption of image 2} Output: Image 2. InstructPix2Pix Instruction: Swap sunflowers

with roses .

LeafInstruct (Ours)

Instruction: Describe a travel log during a hiking. Highlight the scene of plants, insect, and emphasize the beauty of nature.

Input Context: That beauty of nature in all its glory is captivating, with vibrant colors and delicate forms . An insect seems to be trying to eat the plant .





Output: The colors of nature are truly stunning . A wildflower or plant of some sort, possibly a native species, stands out in the landscape . There was a little ladybug in nature, its red shell contrasting beautifully with the surrounding foliage .



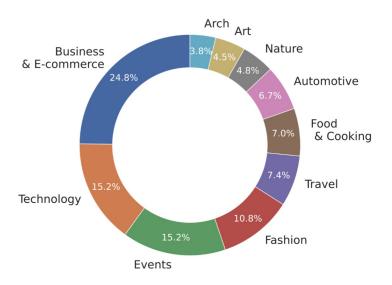


Figure 7: Domain distribution in LeafInstruct.







Main Results of Interleaved Generation

Our approach is highly effective and efficient when it is adapted to existing interleaved generation models.

Model	Text Quality	Perceptual Quality	Image Coherence	TIC	Helpfulness
Proprietary Models					
Gemini1.5 + SDXL	3.37	4.34	3.34	3.98	3.28
GPT-40 + DALL·E 3	3.16	4.44	3.13	4.39	3.46
Open-Source Models					
MiniGPT-5	1.31	3.44	2.06	2.66	1.76
GILL	1.44	4.02	2.12	2.69	1.53
Emu2	1.33	2.29	1.71	1.22	1.87
Chameleon	3.33	0.67	0.28	0.47	1.43
Emu2 + MoSS (Ours)	2.61 (+96.2%)	3.62 (+58.1%)	3.41 (+99.4%)	3.54 (+190.2%)	2.71 (+44.9%)
Chameleon + MoSS (Ours)	2.98 (-10.5%)	2.25 (+235.8%)	1.05 (+275%)	1.7 (+261.7%)	1.82 (+27.3%)



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Qualitative Results

- Our approach demonstrates better helpfulness and instruction-following capabilities.
- Text quality is commonly poor in previous baselines.
- Baselines' generated images are not consistent with inputs.

MiniGPT-5 Input

Output:



Instruction: Describe a team's success in University Rover Challenge. Context: During the challenge, the PCz Team placed the second .



Instruction: Complete a

the countryside for the

weekend .

travel story. Context: I visited

Instruction: Tell me what should do next to plan vegan healthy meals. Context: I've done: I had oatmeal with fruit and a tofu scramble for breakfast in the morning .







Output:

Output: After you eat breakfast make sure to have enough food for dinner .



Output: the gardens

GILL



Output: The team's rover is an autonomous vehicle that can across the surface of the moon.



Output: the tofu

scramble



Output: I was so excited to visit the country side for the

Emu2

weekend . I was so excited to visit the country side for the weekend.



Output: Describe a team's success in University Rover Challenge .



Output: Eat vegan cereal with a non-dairy milk for a quick breakfast .



Emu2+MoSS (Ours) Output: I went to the

countryside and saw many beautiful flowers . I also saw many beautiful buildings .



Output: 16 teams from 10 countries competed in the University Rover Challenge.



Output: Have a good vegan meal for lunch .





Conclusion

- We introduce MoSS, a novel framework that efficiently adapts VLGs by using modality-specialized architectures and parameters for each modality.
- We created LeafInstruct, the first open-sourced large-scale interleaved instruction tuning dataset with 184K high-quality instances spanning more than 10 domains.
- MoSS achieves state-of-the-art performance, significantly surpassing baseline
 VLGs in complex interleaved generation tasks.