



SlideCoder: Layout-aware RAG-enhanced Hierarchical Slide **Generation from Design**

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- Natural language cannot fully express complex layouts
- Multimodal models struggle with dense visual structures
- Generated code often fails to execute correctly

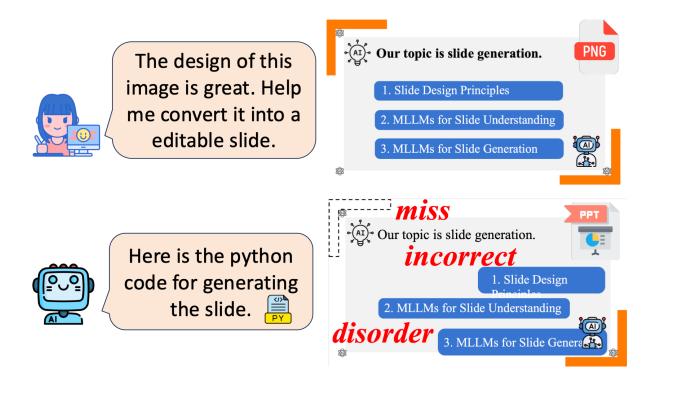


Figure 1: Illustration of slide generation scenarios from design and mistakes made by MLLMs.



- A new benchmark for image-to-slide generation
- Categorized by Slide Complexity Metric (SCM)
- 900 samples across simple, medium, and complex layouts

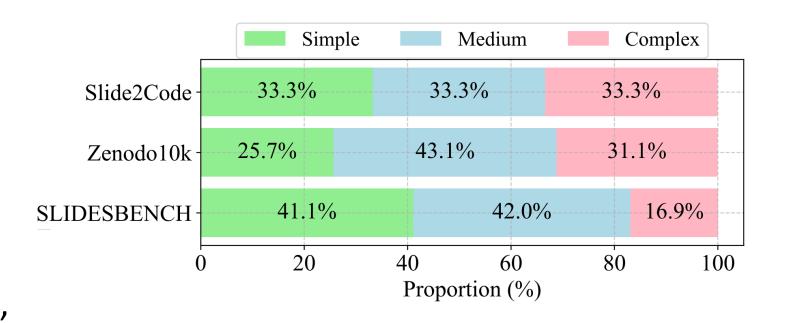
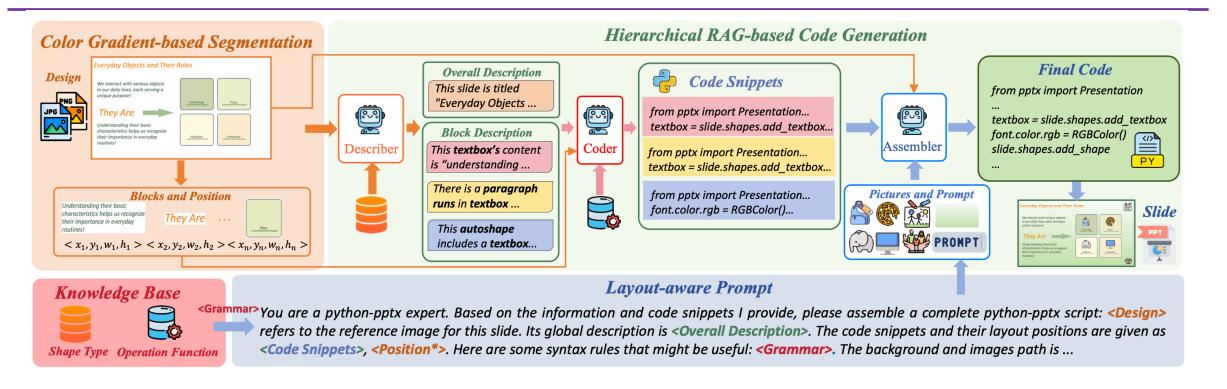


Figure 2: Proportion of samples across three levels in the Slide2Code, Zenodo10k, and SLIDEBENCH datasets.







- CGSeg Color Gradient-based Segmentation
- H-RAG Hierarchical Retrieval-Augmented Generation
- LAP Layout-aware Prompting





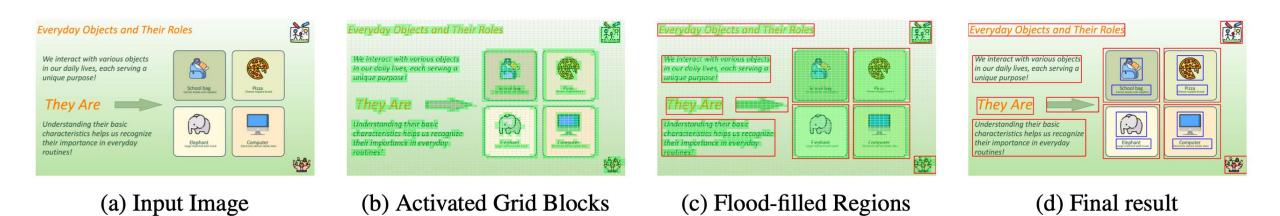
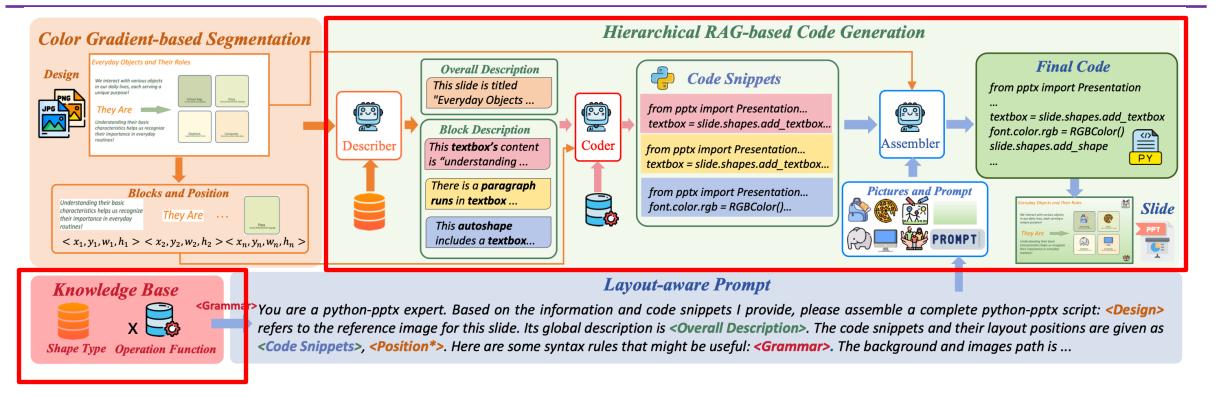


Figure 4: An example of CGSeg applied to a slide reference image. The algorithm begins by computing color gradients (a-b), fills them (c), and recursively segments sub-regions (d).

- Recursive segmentation using color gradient
- Preserves spatial hierarchy and object boundaries
- Produces semantic sub-regions for code generation





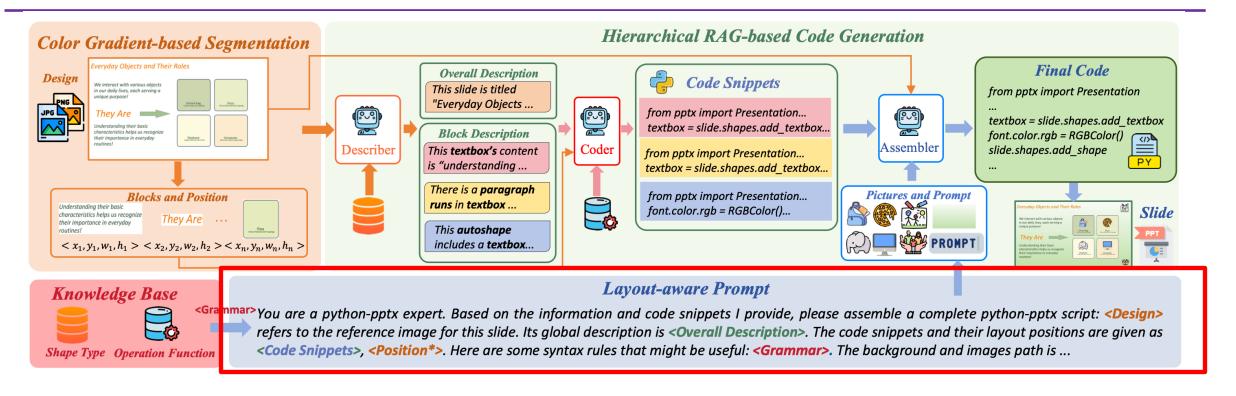


- Two-level knowledge bases:
 - Shape Type KB: Object definitions and templates
 - Operation KB: Functions and syntax patterns
- Three agents: Describer, Coder, Assembler



Layout-aware prompt





- Incorporates layout parameters (x, y, w, h)
- Uses consistent pptx coordinate units (inches)
- Ensures structural and visual alignment



- Based on Qwen2.5-VL-7B
- Reverse-engineered data generation pipeline
- Expands object and style diversity (10 object types, 44 styles)

Table 2: Object Types and Corresponding Style count

Type Name	Ours	AutoPresent's
title	10	3
textbox	10	5
bullet points	8	5
background color	1	1
image	2	2
placeholder	4	_
freeform	2	_
connector	5	_
table	4	_
triangle	5	_

Experimental Results



- Compared models:

 AutoPresent, GPT-4o,
 Gemini 2.0, SlideMaster
- SlideCoder achieves top scores across all difficulty levels
- +40.5 points improvement over baselines

Table 1: Results on Slide2Code (top) and SLIDESBENCH (bottom) using SlideCoder and AutoPresent with different MLLMs. Green, yellow, and red indicate simple, medium, and complex levels in SlideCoder. **Bolded values** mark the best result per level.

80.3 79.1 73.4 79.0 71.6 64.7 75.6 72.6	48.6 41.4 48.5 44.8 51.5 45.2 45.4
79.1 73.4 79.0 71.6 64.7 75.6 72.6	41.4 48.5 44.8 51.5 45.2 45.4
79.1 73.4 79.0 71.6 64.7 75.6 72.6	41.4 48.5 44.8 51.5 45.2 45.4
73.4 79.0 71.6 64.7 75.6 72.6	48.5 44.8 51.5 45.2 45.4
79.0 71.6 64.7 75.6 72.6	44.8 51.5 45.2 45.4
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64.7 75.6 72.6	45.2 45.4
75.6 72.6	45.4
72.6	
	260
(0.0	36.8
63.3	47.1
91.1	76.7
86.4	61.7
82.8	54.2
90.7	87.0
85.5	76.6
81.2	71.6
91.8	89.1
86.2	85.5
82.6	78.4
<u> </u>	
73.7	65.3
66.0	40.4
70.8	66.9
80.0	68.4
80.0	75.0
80.9	78.8
	63.3 91.1 86.4 82.8 90.7 85.5 81.2 91.8 86.2 82.6 73.7 66.0 70.8 80.0 80.0



Table 3: Overall performance of ablation study.

Setting	Execution %	Overall
SlideCoder	100.0	89.9
	100.0	85.8
	100.0	82.2
w/o Layout	100.0	81.2
	93.9	73.6
	93.9	71.8
w/o CGSeg	75.8	55.4
	51.5	39.6
	69.7	48.4
w/o H-RAG	90.9	80.4
	81.8	69.3
	84.8	70.7
Native Setting	75.8	53.9
	48.5	37.4
	66.7	46.9



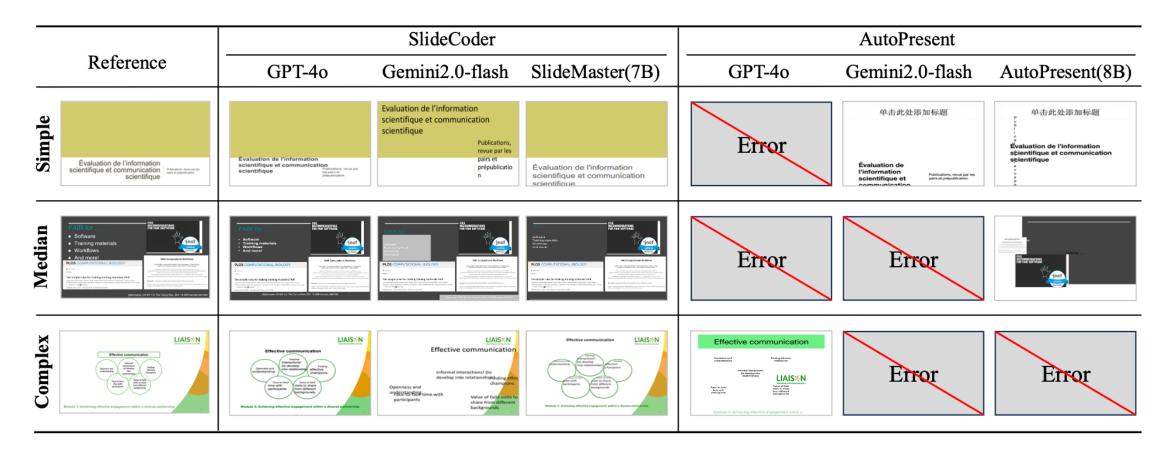


Figure 5: Examples of slides generated by different methods in three difficulty levels.



Thank you!

• Speaker: Wenxin Tang

• Codes: https://github.com/vinsontang1/SlideCoder

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