

CircuitFusion: Multimodal Circuit Representation Learning for Agile Chip Design

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Circuit Representation Learning



Background: AI for IC Design

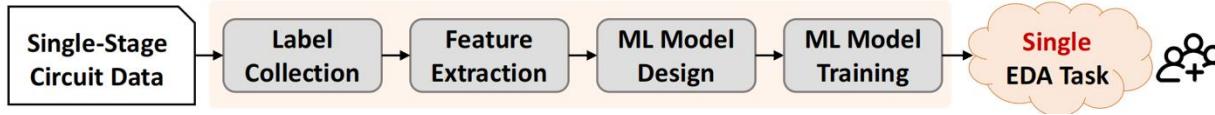
- **Remarkable achievements**
 - **Design quality evaluation**
 - Power, timing, area, routability, etc.
 - **Functional reasoning**
 - Arithmetic word-level abstraction, SAT, etc.
 - **Optimization**
 - Design space exploration, etc.
 - **Generation**
 - RTL code, verification, etc.
 - ...



Paradigm Shift in AI for IC Design

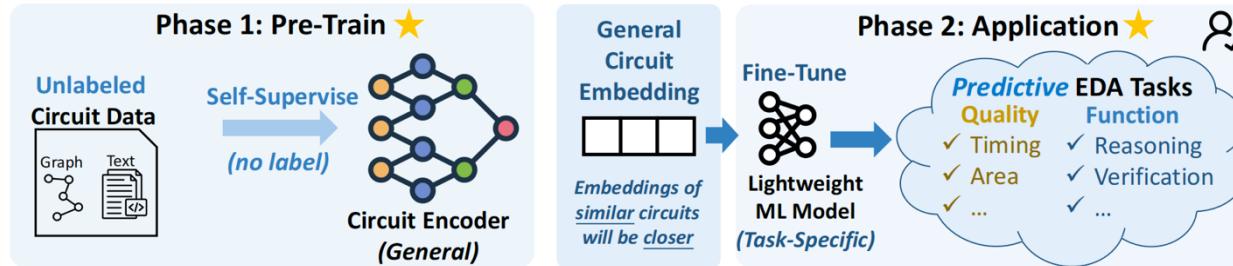
- Traditional: **task-specific supervised learning**

- Tedious, time-consuming, not generalized



- New trend: **general self-supervised representation learning**

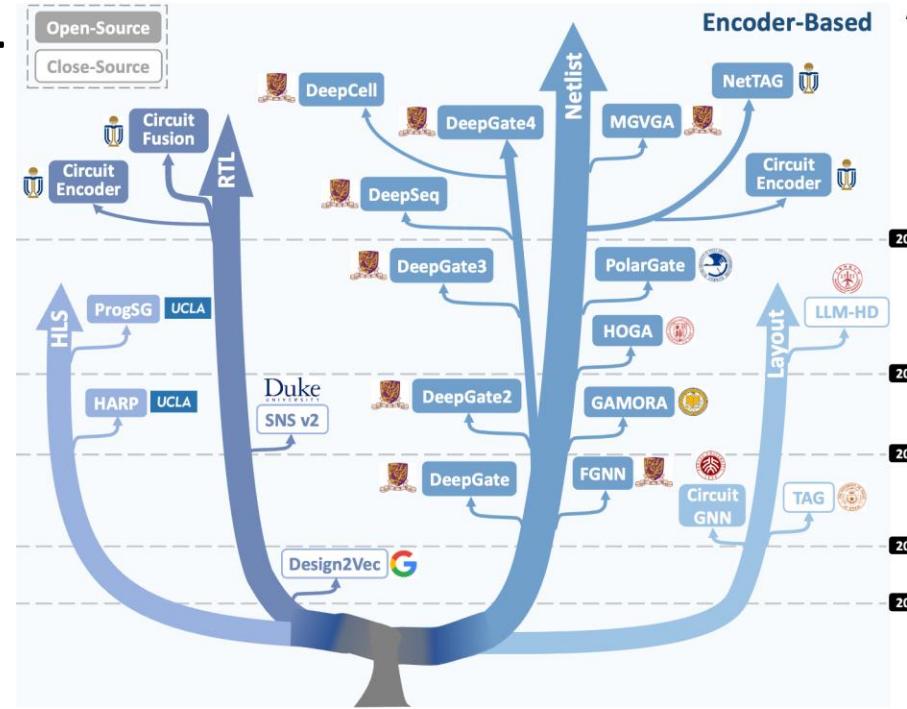
- Encode circuit into **embeddings** for various **EDA tasks**
 - *Pre-training* to capture circuit intrinsics
 - *Fine-tuning* for EDA tasks





Existing Circuit Representation Learning

- Explorations emerging from 2021
- Cover all main design stages
 - HLS, RTL, netlist, layout
- Research focus
 - Pre-training techniques
 - Circuit-related supervisions
 - Circuit-specific self-supervised
 - Predictive EDA tasks
- **Limitation: only focus on circuit *graph structure***





Multimodal Representation Learning

- **Encode & fuse information from diverse modalities**

- Vision-language
- Graph-language
- Software-graph
-



- **Can we fuse multiple circuit modalities to learn better circuit representation?**

CircuitFusion: Multimodal Circuit Representation Learning



Summary of Circuit Modalities

- Multimodal nature of RTL-stage circuits

- ❖ **Functionality Summary**
- ❖ **Implementation Details**

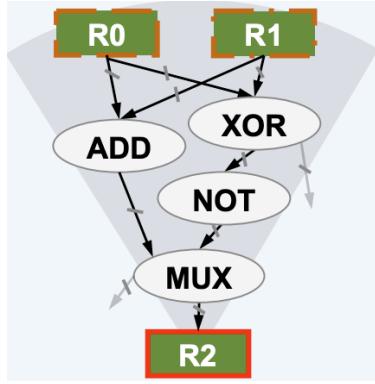
**Functionality
Summary**

semantic

```
reg [1:0] R0,R1;  
reg [2:0] R2;  
wire [2:0] W1,W2;  
...  
assign W1 = R0 + R1;  
...  
always @ (posedge clk)  
R2 <= W2;
```

**HDL
Code**

8



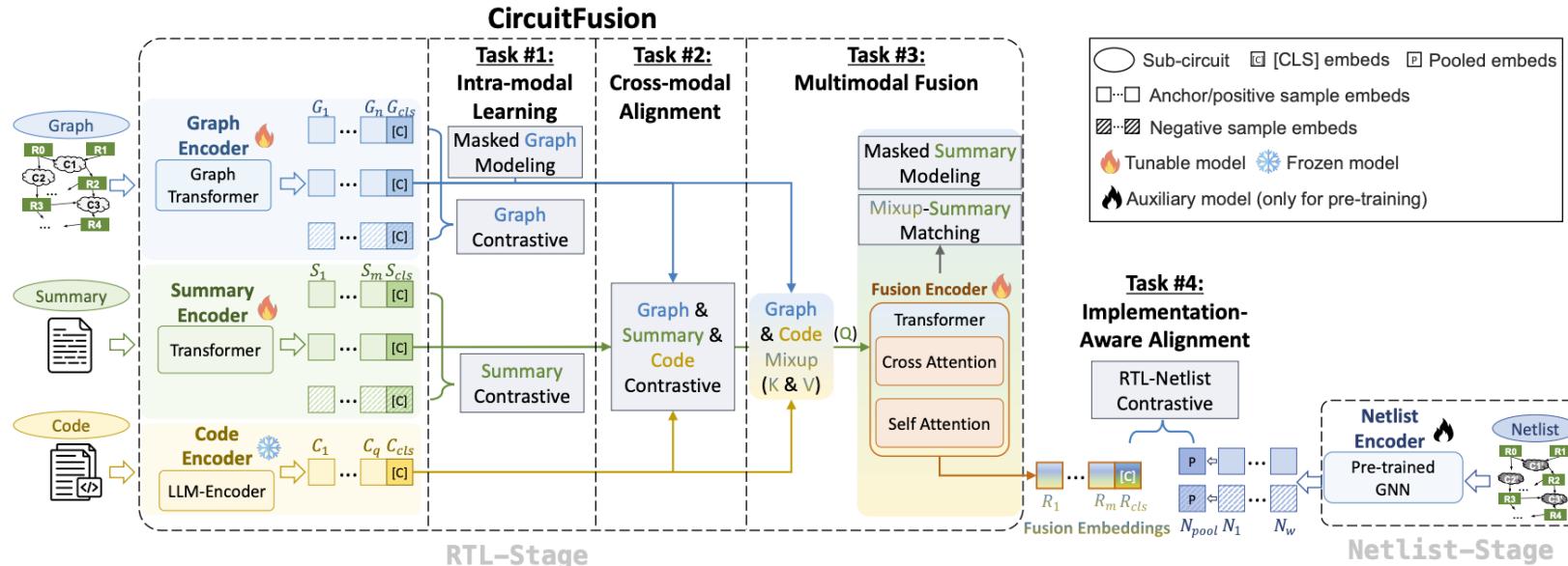
**Structure
Graph**

structure



CircuitFusion Overview

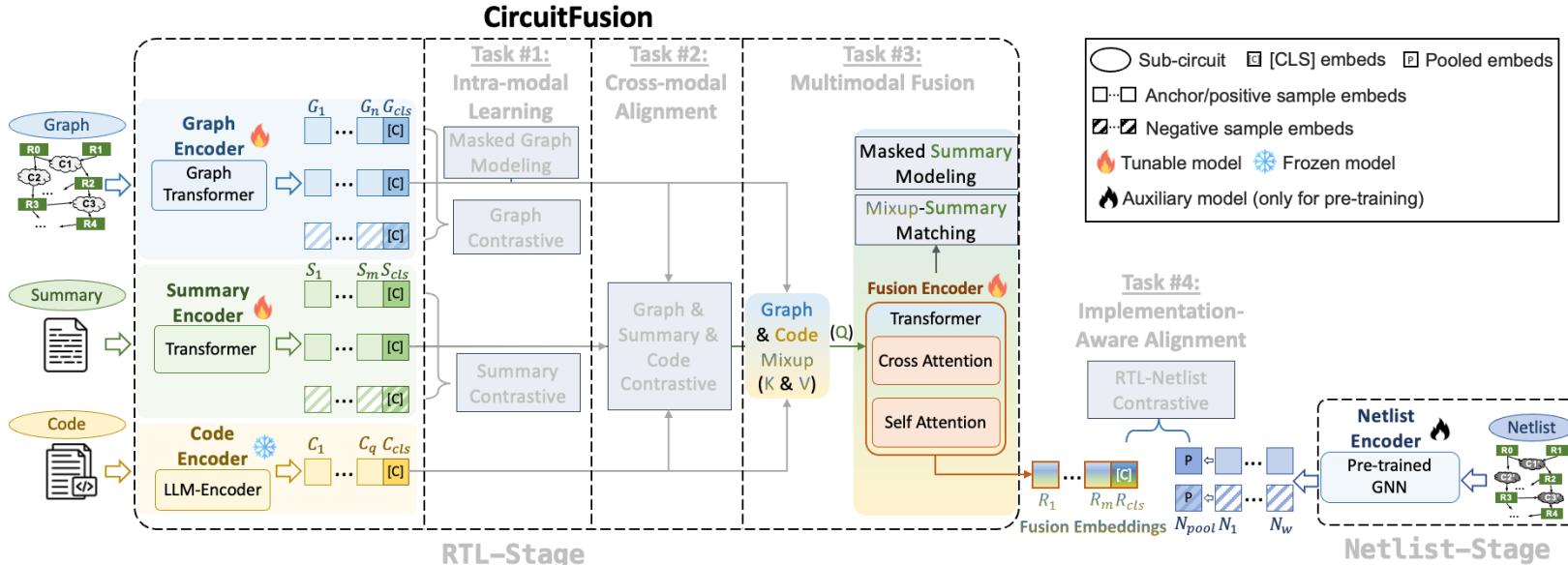
- Multimodal fused & implementation-aware RTL encoder
 - Propose 4 strategies according to 4 unique circuit properties
 - Achieve SOTA performance on various EDA tasks





CircuitFusion Model Architecture

- 3 **unimodal encoders**: graph, summary, code
- 1 **multimodal fusion encoder**: cross attention, summry-centric fusion
- 1 **auxiliary netlist encoder**: implementation-aware





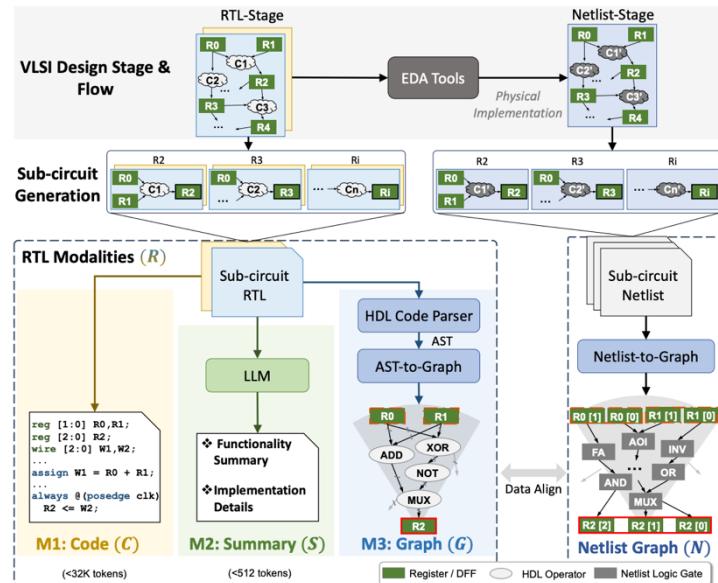
Step 1: Circuit Preprocessing

- **Property 1: parallel execution**

- Combinational logic calculates simultaneously
- Sequential registers are updated only at each clock cycle

- **Strategy 1: sub-circuit generation**

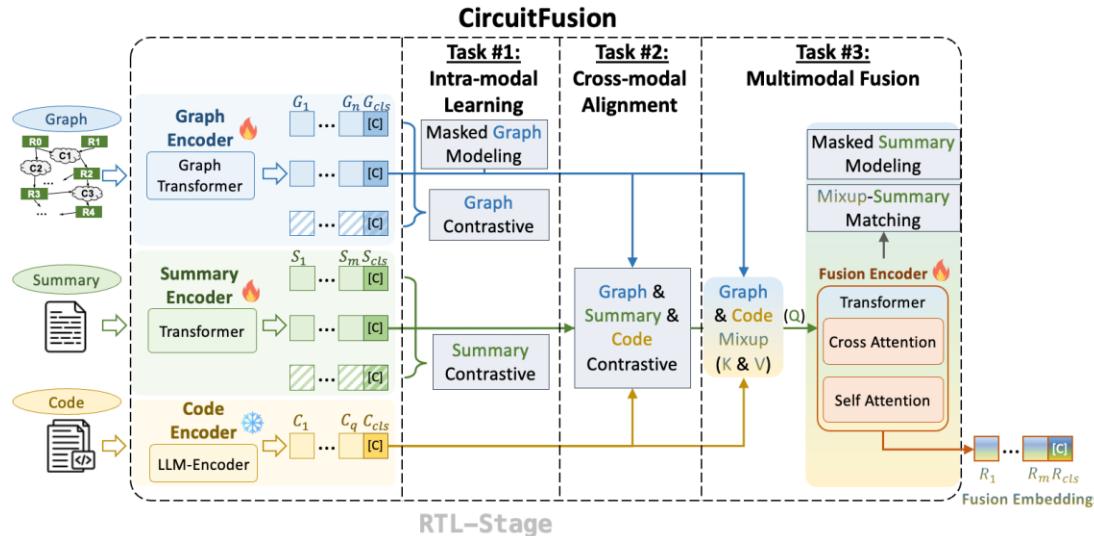
- Split based on register cones
 - Backtrace all combinational input logic
- Advantages
 - Consistency in Modality & stage
 - Complete state transition of 1 cycle
 - Intermediate granularity





Step 2: CircuitFusion Pre-Training

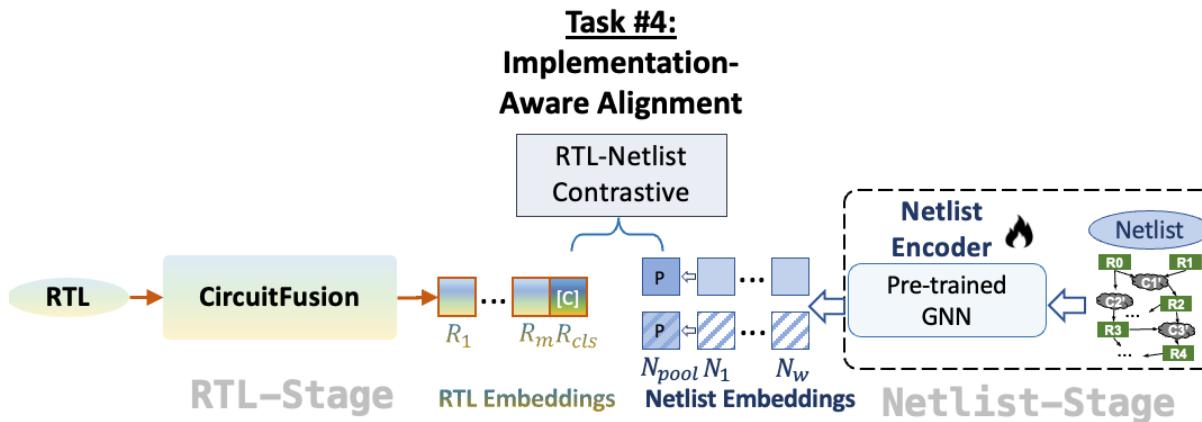
- **Property 2: functional equivalent transformation**
 - Circuit w. similar **function** may have different **structures**
- **Strategy 2: semantic-structure pre-training**
 - Self-supervised Task #1-3 for each modality and multimodal fusion





Step 2: CircuitFusion Pre-Training

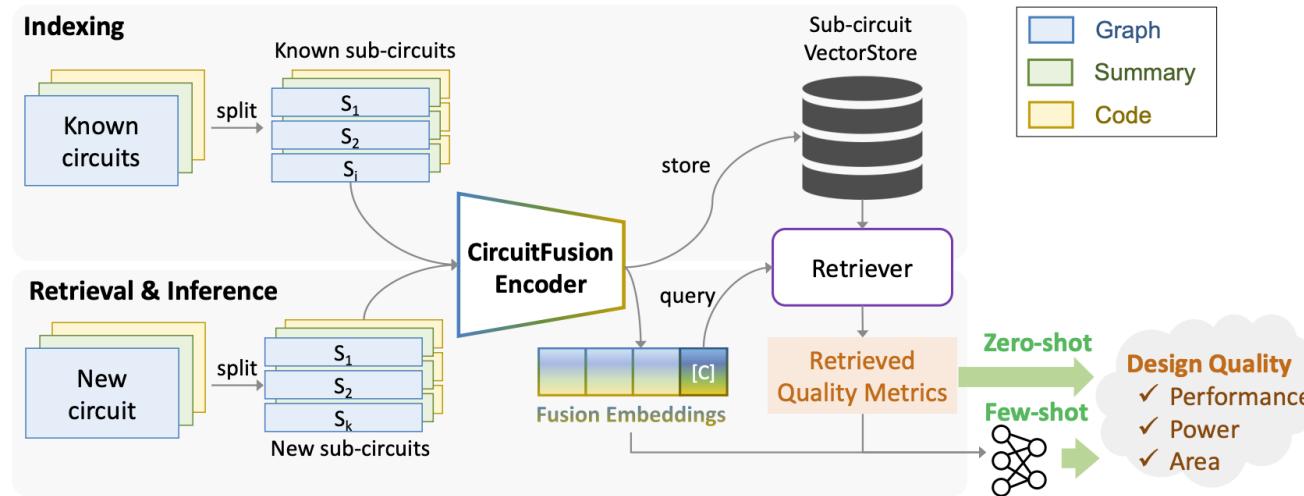
- **Property 3: multiple design stages**
 - RTL (high-level semantics) → netlist (low-level details)
- **Strategy 3: implementation-aware alignment**
 - Pre-training with netlist encoder across design stage (Task #4)





Step 3: Application for EDA Tasks

- **Property 4: circuit reusability**
 - Reuse circuit IPs rather than design from scratch
- **Strategy 4: Retrieval-augmented inference**
 - Retrieves most similar circuits based on embeddings



Experimental Results

Design Quality Prediction Tasks



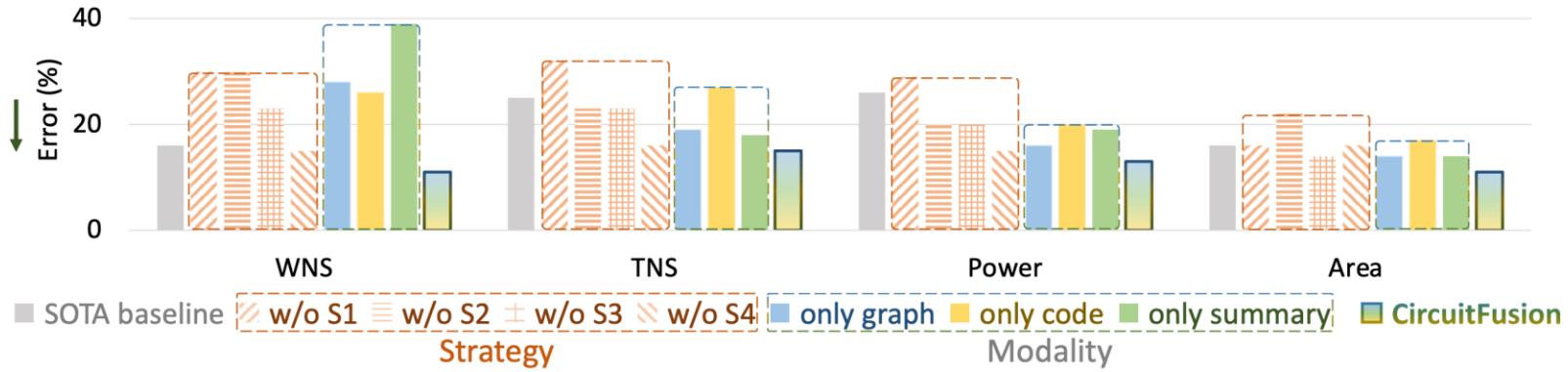
- SOTA performance on RTL-stage PPA prediction vs.
 - Circuit task-specific solutions
 - Text encoders
 - Software code encoders

Type	Method	Slack		WNS		TNS		Power		Area	
		R	MAPE								
Hardware Solution	RTL-Timer	0.85	17%	0.9	16%	0.96	25%	N/A		N/A	
	MasterRTL	N/A		0.89	18%	0.94	28%	0.89	26%	0.98	16%
	SNS v2	N/A		0.82	22%	N/A		0.76	28%	0.93	25%
Text Encoder	NV-Embed-v1	N/A		0.49	17%	0.97	55%	0.85	44%	0.86	24%
Software Code Encoder	UnixCoder	N/A		0.46	21%	0.95	44%	0.83	29%	0.85	26%
	CodeT5+ Encoder	N/A		0.55	21%	0.63	43%	0.49	46%	0.45	39%
	CodeSage	N/A		0.23	25%	0.86	45%	0.8	38%	0.77	41%
Ours	CircuitFusion	0.87	12%	0.91	11%	0.99	15%	0.99	13%	0.99	11%



Ablation Study

- Impact on proposed **strategies** and circuit modalities





Design Quality Prediction Tasks

- Zero-shot retrieval
 - First hardware solution to support zero-shot inference
 - Outperform text / software encoders

Table 3: MAPE(%) results of the zero-shot top-k similar circuit retrieval.

Method	Slack				Sub-circuit Power				Sub-circuit Area			
	top-1	top-3	top-5	top-10	top-1	top-3	top-5	top-10	top-1	top-3	top-5	top-10
LLM Encoder	51	35	33	34	92	90	90	90	90	88	88	88
UnixCoder	56	36	36	36	90	89	90	91	89	88	89	89
CodeT5+ Embedding	57	35	35	36	88	87	89	90	87	86	87	88
CodeSage	50	36	36	36	89	87	88	91	88	85	86	87
Ours	21	22	23	26	36	40	42	53	35	40	42	51

- Performance scaling up w. model & data size



Conclusion & Future Work



Conclusion

- **CircuitFusion:** first multimodal RTL encoder
 - 4 strategies according to 4 unique circuit properties
 - Support various EDA tasks
- Future work
 - Multimodal netlist encoder via text-attributed graph [DAC'25]
 - Align circuit encoders with generative LLM decoders



Thank You!