Circuit Transformer: A Transformer That Preserves Logical Equivalence

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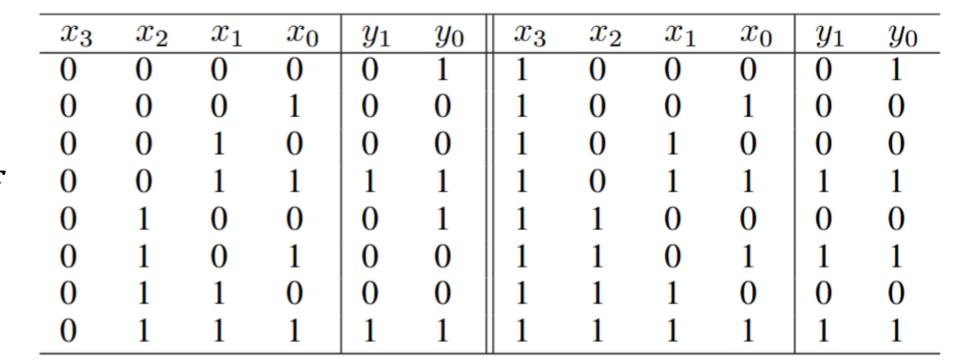
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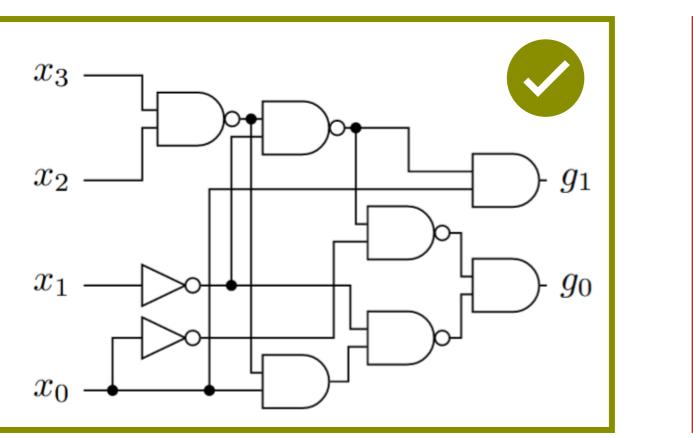
Problem Setting & Challenges

Implementing Boolean functions with logic circuits (A fundamental problem in digital design!)

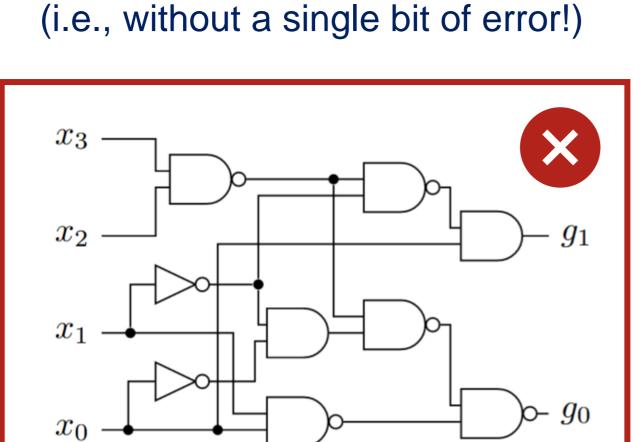
Boolean function f



Logic circuit g



Outputs match for all 16 possible inputs



Must be **exactly** equivalent

Outputs not exactly match (15 match, 1 not match)

Question: Can generative neural models directly generate strictly equivalent circuits for a Boolean function?

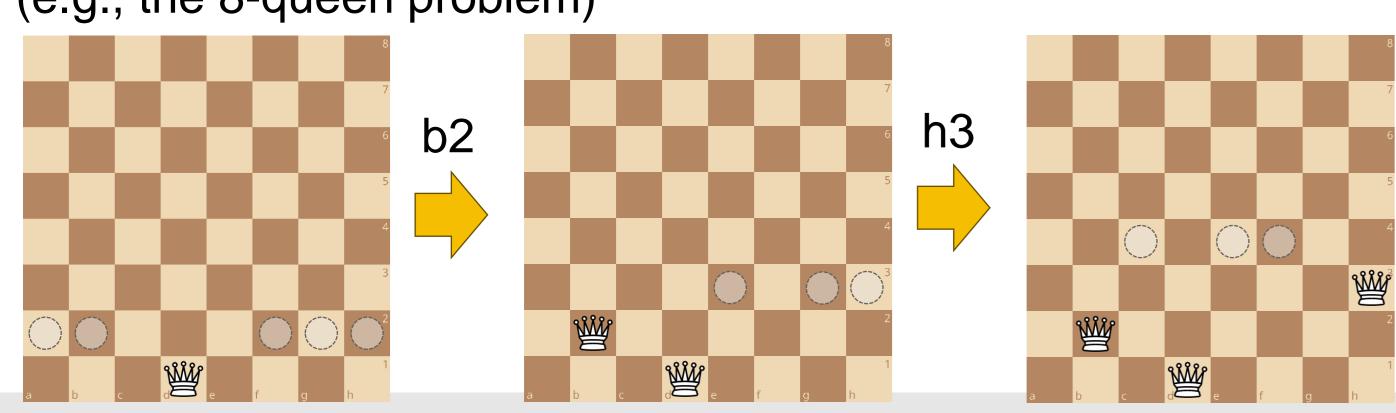
Typical answer is No (generative models occasionally make mistakes). Previous AI approaches strengthens traditional symbolic methods.

Motivation

Valid choices:

a2,b2,f2,g2,h2

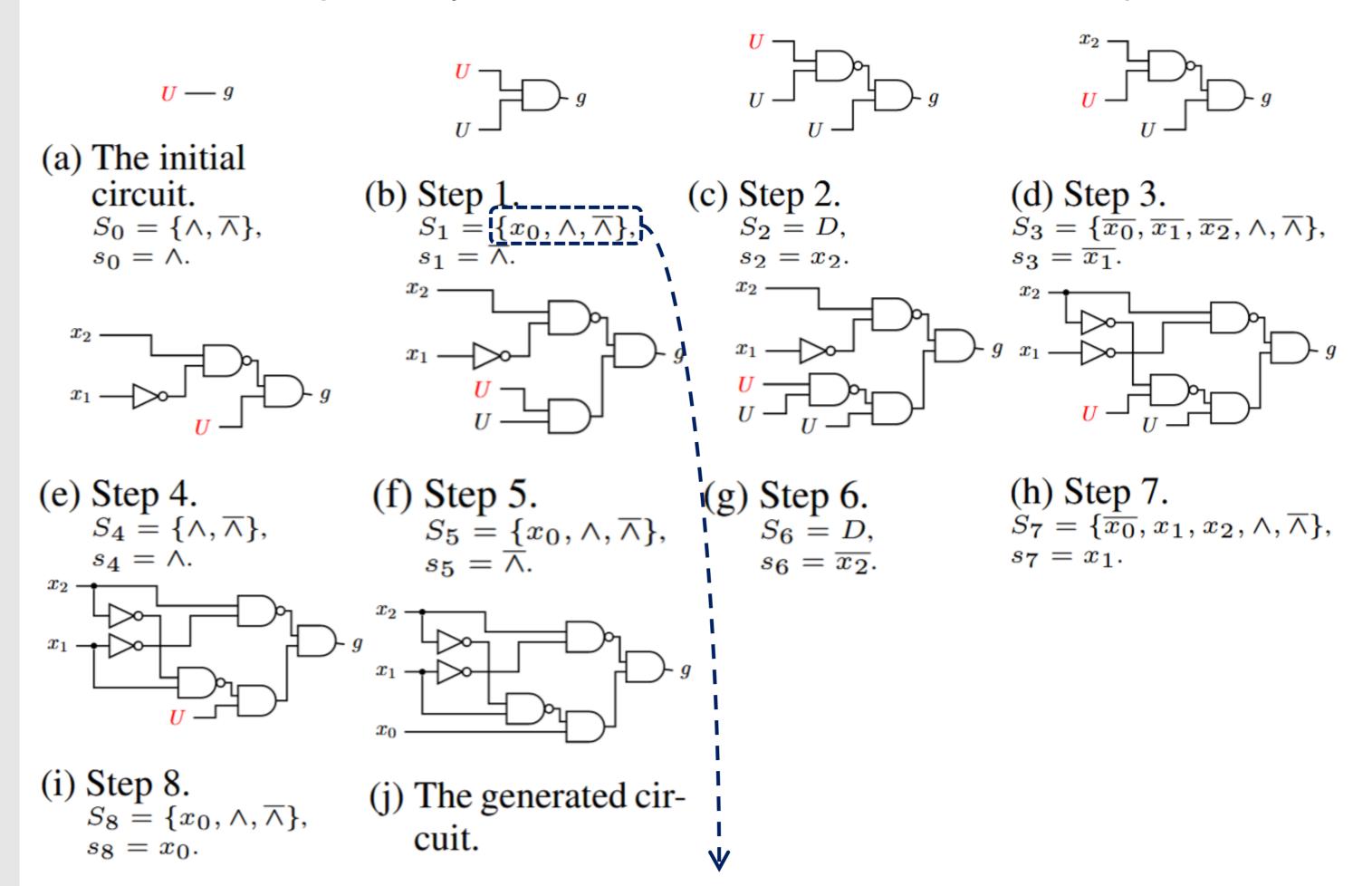
Complex constraints can be satisfied step-by-step! (e.g., the 8-queen problem)



Our Method

Generate circuits step-by-step with equivalence preserved

- Construct from outputs to inputs, start from a wildcard node U
- In each step, replace a wildcard node U with a new gate / input
- In each step t, only choose from the valid choices S_t



How to find valid choices in each step

Valid choices for U: $x_0, \Lambda, \overline{\Lambda}$

U = Unknown

Red = Conflict with f

		x_0	f	The value of g' when U is replaced by							
x_2	x_1			x_0	$\overline{x_0}$	x_1	$\overline{x_1}$	x_2	$\overline{x_2}$	Λ	$\overline{\wedge}$
0	0	0	0	0	U	0	U	0	U	U	U
0	0	1	1	U	0	0	U	0	U	U	U
0	1	0	0	0	U	U	0	0	U	U	U
0	1	1	0	U	0	U	0	0	U	U	U
1	0	0	0	0	U	0	U	U	0	U	U
1	0	1	0	U	0	0	U	U	0	U	U
1	1	0	0	0	U	U	0	U	0	U	U
1	1	1	1	U	0	U	0	U	0	U	U

Replace *U* with each

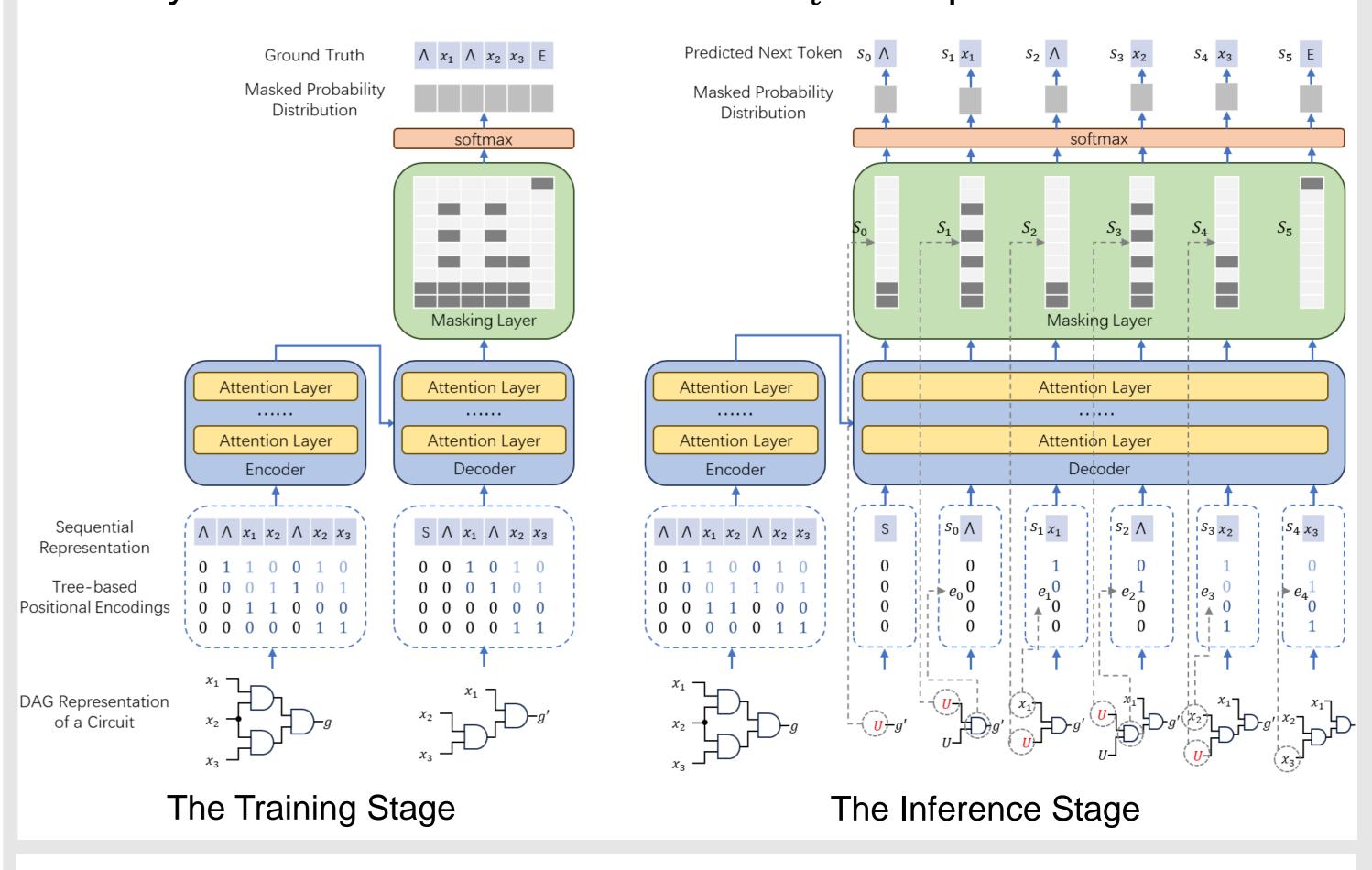
conflict occurs

possible input / gate to see

whether any equivalence

The Circuit Transformer

- The Boolean function is encoded by the Transformer encoder
- A masking layer is added before the softmax layer of the Transformer decoder
- Only allows tokens in valid choices S_t to be predicted

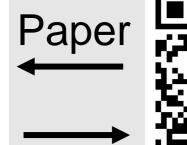


Experiments

Train a Circuit Transformer with 88M parameters to generate minimized circuits for 8-input, 2-output Boolean functions.

	In distribut	ion	Out of distribution					
Methods	Random circ	uits	IWLS FFWs					
Methods	Unsuccessful cases	Avg. size	Unsuccessful cases	Avg. size				
Boolean Chain	5.07% (5.07%)	15.25	11.36% (11.26%)	17.24				
Boolean Chain (beam size $= 16$)	2.16% (2.16%)	14.89	6.34% (6.29%)	17.15				
Boolean Chain (beam size $= 128$)	1.91% (1.91%)	14.87	5.97% (5.94%)	17.15				
AIGER	4.32% (4.32%)	15.14	8.35% (7.77%)	17.19				
AIGER (beam size $= 16$)	1.85% (1.85%)	14.87	4.62% (4.37%)	17.12				
AIGER (beam size $= 128$)	1.71% (1.71%)	14.86	4.24% (3.99%)	17.12				
Circuit Transformer w/o TPE	2.14% (0%)	15.02	6.63% (0%)	17.33				
Circuit Transformer	1.14% (0%)	14.79	4.76% (0%)	17.17				
Circuit Transformer ($K = 10$)	0.20% (0%)	14.02	2.83% (0%)	16.92				
Circuit Transformer ($K = 100$)	0.17% (0%)	13.73	2.63% (0%)	16.73				
Resyn2 (ground truth for training)	/	14.56		16.82				
Better performance with	Zero violation of							
MCTS enabled	equivalence constraints							







Slide, Poster, Video and Online Demo: https://snowkylin.github.io/publications

Install: pip install circuit-transformer Correspondence email: xihan.li@cs.ucl.ac.uk





