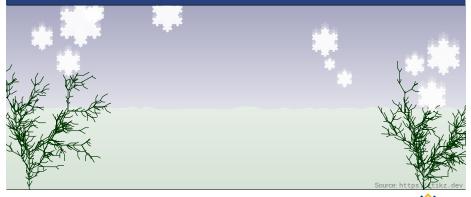
DeTikZify



DETIKZIFY: Synthesizing Graphics Programs for Scientific Figures and Sketches with TikZ

Jonas Belouadi and Simone Ponzetto and Steffen Eger Natural Language Learning Group (NLLG) & Data and Web Science Group (DWS)



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On Scientific Figures



Observation 1

Even though sketching ideas on paper is easy, creating high-quality scientific figures can be time-consuming and challenging.

Source: https://www.britannica.com/biography/Pythagoras



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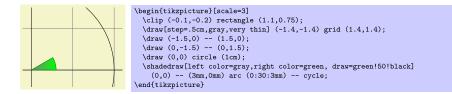
Observation 2

Recreating existing figures that are stored in low-level formats (JPG, PNG, PDF, SVG, ...) which do not preserve semantic information is equally complex.

Source: https://www.britannica.com/biography/Pythagoras

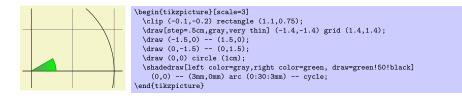


We introduce DETIKZIFY, a multimodal LLM that automatically synthesizes scientific figures based on sketches and existing figures as semanticspreserving TikZ graphics programs. We leverage TikZ because it is:



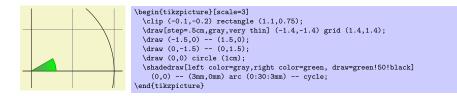


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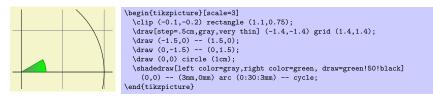


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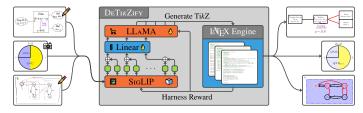


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Key Contributions

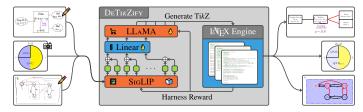


datasets We collect DATIKZ_{V2} with over 360k TikZ graphics; SKETCHFIG consisting of sketches paired with scientific figures; and METAFIG, a meta-dataset of figures and associated metadata.





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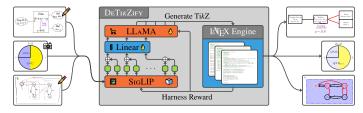


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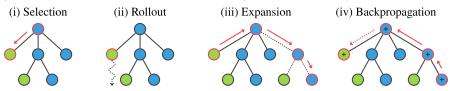
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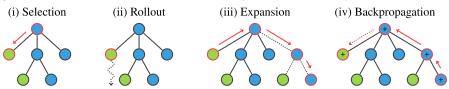


The algorithm iteratively builds a search tree and repeatedly runs simulations. Each node's state consists of n lines of TikZ code, and edges represent possible continuations. Each simulation performs four steps:





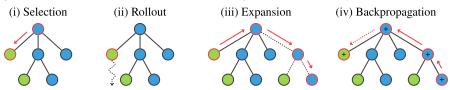
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(i) Selection: Select a path to a leaf node based on the scores of each node.



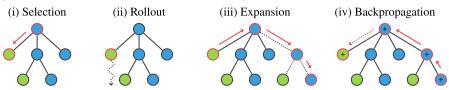
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(ii) Rollout: Starting from the state of the leaf node, use DETIKZIFY as a rollout policy and complete the code until an <eos> token is reached.



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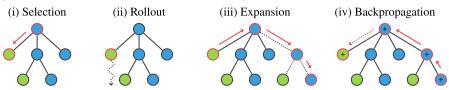
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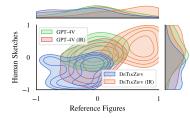


(i) Selection: Select a path to a leaf node based on the scores of each node.

- (ii) Rollout: Starting from the state of the leaf node, use DETIKZIFY as a rollout policy and complete the code until an <eos> token is reached.
- (iii) Expansion: Expand the search tree by appending nodes from the rollout.
- (iv) Backpropagation: Use DETIKZIFY to evaluate the rollout and backpropagate the reward score to each node until the root is reached.



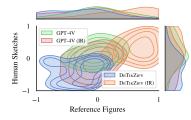
Evaluation



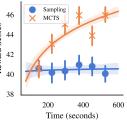
- At first, humans rate GPT-4V outputs higher than DETIKZIFY from -1 (bad) to 1 (good).
- However, after applying IR with MCTS, DETIKZIFY ranks highest. Further, IR techniques do not work well with GPT-4.



Evaluation

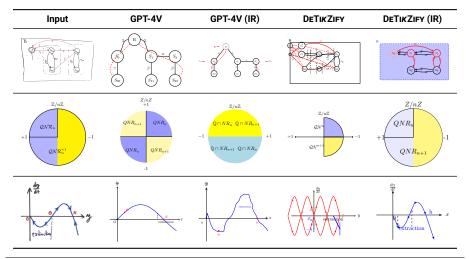


- At first, humans rate GPT-4V outputs higher than DETIKZIFY from -1 (bad) to 1 (good).
- However, after applying IR with MCTS, DETIKZIFY ranks highest. Further, IR techniques do not work well with GPT-4.
- With MCTS, DETIKZIFY consistently improves over time, and even after 10 minutes does not appear to converge.
- Standard sampling does not lead to better scores over time due to the absence of a feedback loop.





Examples





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Interested? There's more!



In our paper, we explore various building blocks of DETIKZIFY and additional baselines. We assess efficiency, code similarity, and image similarity through automatic evaluation, and correlate them with humans. We also examine the quality of synthetic sketches and show that our models are unaffected by training data memorization.

Paper https://arxiv.org/abs/2405.15306 Code https://github.com/potamides/DeTikZify Artifacts https://hf.co/collections/nllg/ detikzify-664460c521aa7c2880095a8b Demo https://hf.co/spaces/nllg/DeTikZify Group https://nl2g.github.io



