

Tuning the Learning of Circuit-Based Classifiers

Master thesis

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Motivation

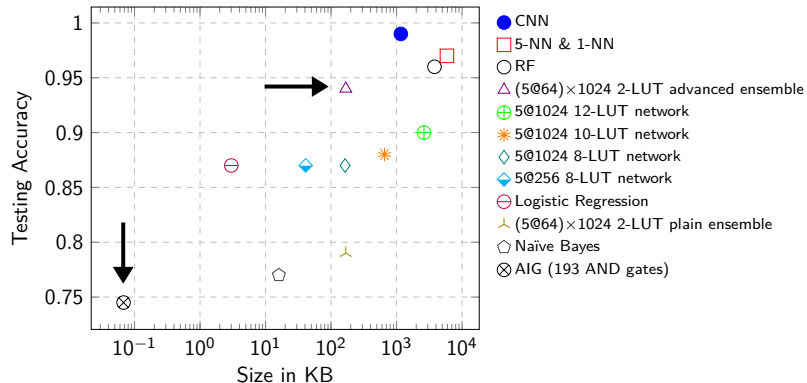
- ▶ Neural Networks blow our mind, but often have lots of parameters
- ▶ High computational costs for training and inference
 - ▶ Environmental impact is a concern [1]
 - ▶ Need to deploy on cloud (and thus, internet connection)
- ▶ Circuit-based models could be a step toward efficiency
- ▶ Construct classifiers that work internally with binary variables

Approach

- ▶ Most common: gradually shrink neural architectures [2]
- ▶ We already start at binary
- ▶ Lookup table (LUT) networks [3], [4] and LUT learning algorithm
- ▶ And-inverter graphs (AIGs) and local search

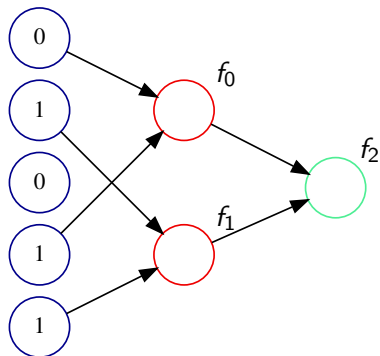
Results

Model performance on Binary-MNIST



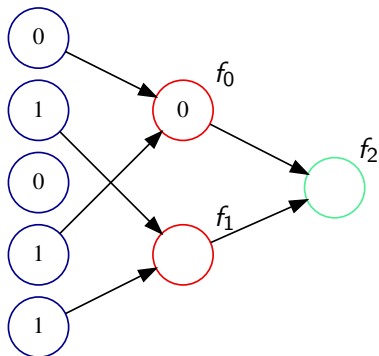
Simple LUT Network

bit-pattern	f_0	f_1	f_2
00	1	1	1
01	0	0	1
10	1	0	0
11	0	1	0



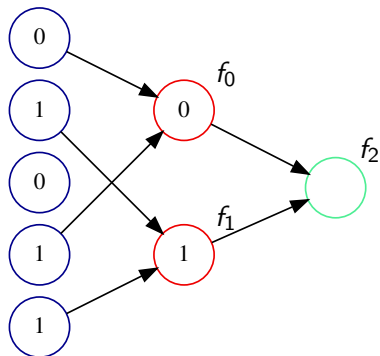
Simple LUT Network

bit-pattern	f_0	f_1	f_2
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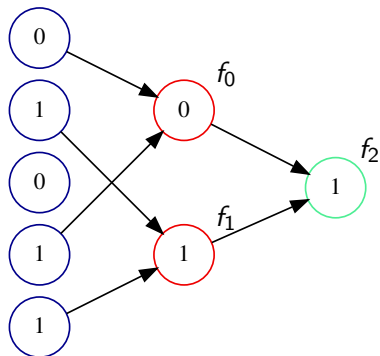
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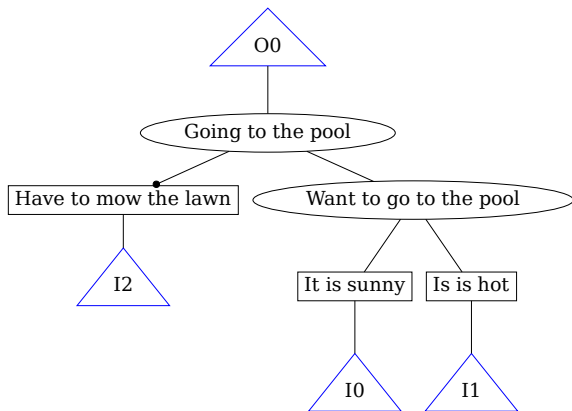
LUT networks: Our Contributions

- ▶ Wrote Python code from scratch in approximately 600 LOC
 - ▶ Used NumPy library extensively [5]
- ▶ Recreated one experiment from [3], got the same results
- ▶ Taking the idea of LUT networks further:
 - ▶ Modify existing LUT networks or the LUT network learning algorithm
 - ▶ Enhance the dataset using feature engineering
 - ▶ Combine many LUT networks of low arity (ensembling)

LUT networks: Findings

- ▶ Small tweaks boosted the performance of low-arity LUT networks
- ▶ Feature engineering using Sobel-inspired operator gave minor improvement
- ▶ AdaBoost.M1-inspired ensemble had reduced size and gave best accuracy yet (90% → 94%)

AIG Classifier



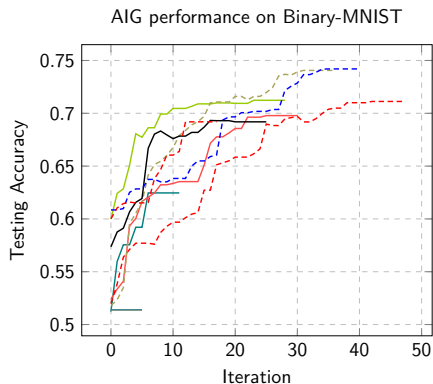
$$(\neg I2) \wedge (I0 \wedge I1)$$

AIG Classifier: Our Contributions

- ▶ Usually, AIGs used as intermediate representations for circuits
- ▶ Our idea: use AIGs as classifiers
- ▶ Devised an initialization scheme and a local search algorithm
- ▶ Wrote C++ code from scratch in approximately 900 LOC
 - ▶ AIG structure, local search, export to other formats
 - ▶ AIGER library [6] used for inference

AIG Classifier: Results

- ▶ We conducted 42 experiments
- ▶ The resulting AIGs were tiny
- ▶ Implementation on edge devices is promising



Conclusion

- ▶ Topic of this thesis: building binary predictors and starting from binary
- ▶ LUT: Recreated paper [3], got the same results
- ▶ Improved LUT network accuracy while reducing model size
- ▶ AIG: Devised initialization and local search scheme
- ▶ Tiny AIG size is promising for implementation on edge devices

References I

- [1] R. Schwartz, J. Dodge, N. A. Smith, and O. Etzioni, “Green AI,” *Communications of the ACM*, vol. 63, no. 12, pp. 54–63, 2020.
- [2] D. Liu, H. Kong, X. Luo, W. Liu, and R. Subramaniam, “Bringing AI to edge: From deep learning’s perspective,” *Neurocomputing*, 2021.
- [3] S. Chatterjee, “Learning and memorization,” in *International conference on machine learning*, 2018, pp. 755–763.
- [4] S. Chatterjee and A. Mishchenko, “Circuit-based intrinsic methods to detect overfitting,” in *Proceedings of the 37th international conference on machine learning, ICML 2020, 13-18 july 2020, virtual event, 2020*, vol. 119, pp. 1459–1468. Available: <http://proceedings.mlr.press/v119/chatterjee20a.html>

References II

- [5] C. R. Harris *et al.*, “Array programming with NumPy,” *Nature*, vol. 585, no. 7825, pp. 357–362, Sep. 2020, doi: 10.1038/s41586-020-2649-2.
- [6] A. Biere, “The AIGER And-Inverter Graph (AIG) format version 20071012,” Institute for Formal Models; Verification, Johannes Kepler University, Altenbergerstr. 69, 4040 Linz, Austria, 07/1, 2007.