

LiCE

Generating Likely Counterfactuals using Sum-Product Networks

Jiří Němeček, Tomáš Pevný, Jakub Mareček

nemecek.jiri@fel.cvut.cz Artificial Intelligence Center Faculty of Electrical Engineering Czech Technical University in Prague

Counterfactual Explanations (CEs)

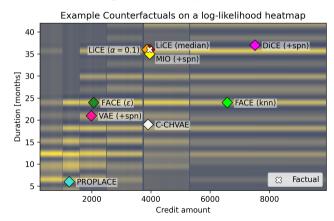


- Minimal change of input to achieve desired output.
- Example: "You would get the loan if your income was \$100,000 instead of \$75,000."
- Many desiderata for CEs [1].
 - Plausibility: The counterfactual should have a high likelihood w.r.t. the data distribution.
- A trade-off between plausibility and similarity [2].

LiCE



- Successfuly optimize of plausibility and similarity at once.
 - Under further constraints (e.g., actionability, discrete data).



Plausibility in LiCE

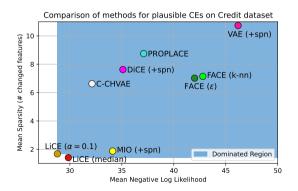


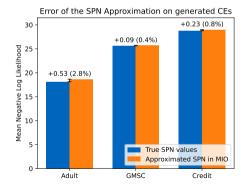
- Utilizing trained Sum-Product Networks:
 - DAG, computing sample likelihood.
 - Leaves are (univariate) distributions.
 - Inner nodes are
 - weighted sums (mixtures)
 - or **products** of children.
 - They model both discrete and real-valued random variables.
- We formulate them in **Mixed-Integer Linear Optimization**.
- And solve with a constraint on counterfactual's likelihood.
 - Or optimize a linear combination of similarity and plausibility.

Conclusion



- LiCE outperforms other methods for plausible CEs on likelihood and similarity.
- We tightly approximate SPNs within MIO, opening up various future directions.





Thank you



• Please contact me with any questions at:

contact@nemecekjiri.cz

• LiCE implementation is available at:

github.com/Epanemu/LiCE

References



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Keep Your Friends Close and Your Counterfactuals Closer: Improved Learning

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