Utilization of GAN for Automatic Evaluation of Counterfactuals: Challenges and Opportunities*

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Abstract. Over the past few years, Explainable Artificial Intelligence (XAI) has grown significantly due to the fact that successful deep learning models are still difficult to understand and interpret. XAI aims to enable better interpretability of the classifications made by the neural networks for humans. In XAI research, counterfactual explanations are proven to be very effective in explaining the model's mistakes, describing what changes should be applied to a particular input (image in our case) to attain the correct classification [1]. However, systematic evaluation of counterfactuals is challenging and requires substantial human input. In this work we focus on evaluating semantic textual explanations (expressed as attribute-value pairs) on birds classification (CUB-200-2011 [2] dataset). Being textual and not visual, the explanations are hard to systematically validate through the CNN. To tackle this problem, we experimented on the use of Generative Adversarial Networks (GANs) to modify a misclassified image based on a textual counterfactual. The resulting counterfactual image generated affected the original misclassification outcome but not as strongly as we expected. We believe this could be due to the known problem of poor resolution in GAN-generated images, and the challenge of multiple attribute modifications. This paper reports on the challenges of using GANs to systematically assess the quality of textual counterfactuals via counterfactual image

Keywords: Explainable AI (XAI) \cdot Counterfactual explanations \cdot Generative Adversarial Networks (GAN) \cdot Computer Vision.

References

generation.

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