

and suspect that this can make the resulting interpretability claims unreliable. Future work should examine this hypothesis closely and furnish further demonstrations. Finally, it is interesting to develop more principled techniques for activation patching or propose other methods for localization.

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