

Dividing and Conquering a BlackBox to a Mixture of Interpretable Models:

Route, Interpret, Repeat

TLDR: Extracting a mixture of interpretable models from a BlackBox to provide instance specific concept-based explanations using First-order logic (FOL).



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LAB

Post hoc explanation

BATMAN

Pros

· Does not alter the Black box.

Cons

- Inconsistent explanations.
- No recourse.

Interpretable by design

Support concept intervention.

Cons

- Harder to train.
- Sub par performance.

How to blur this gap?

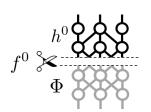
Desirable properties

- Does compromise the performance.
- Can be intervened to fix the misclassification

Design choices

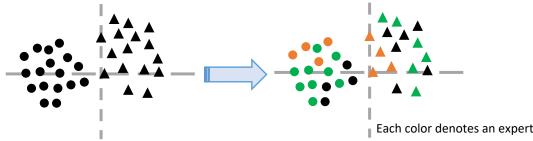
- Carve interpretable models from Blackbox.
- Concept based
- First order logic for concept interaction

Assumptions





Carve out interpretable models from Black box



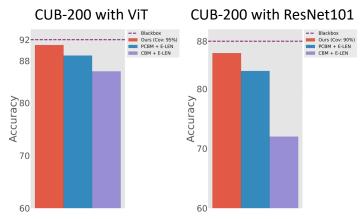
Route Interpret Repeat Fix Φ , update hResidual (r^0) Residual (r^2) Residual (r^3) Residual (r^1) Blackbox Model Symbolic Model Selector * SelectiveNet [Geifman et al.] optimization Each g is E-LEN [Barberio et al.], constructing FOL * Continue till at least 90% samples covered **Capturing heterogenous explanations**

* Extracted from ViT-based BlackBox

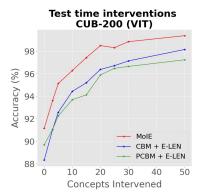


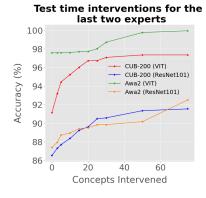


Not compromising performance



Test time interventions





Also in our paper,

- + we experiment with a diverse set of datasets and architectures
- + we achieve higher concept completeness scores
- + VIT-based experts compose less concepts than CNN-based
- + we eliminate shortcut learning problem (SCIS w)
- + we efficiently transfer the experts to new domain (IMLH w)

