

# Improving Machine Learning Explainability with a Graphical User Interface

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### Improving Machine Learning Explainability with a Graphical User Interface

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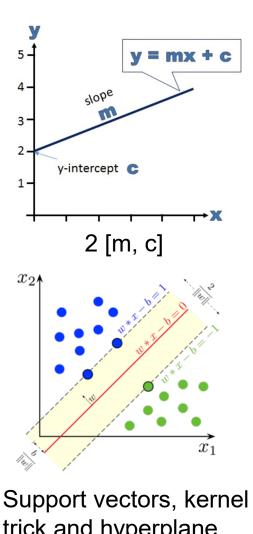
#### Improving Machine Learning Explainability with a Graphical User Interface

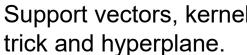


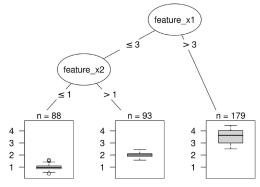
Explainability comes in different forms depending on which model you use.

- How do we explain these models to the user?
- How much would you have to explain to go from an input to an output?

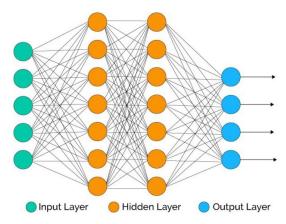
Hyperparameters Post Model Data & Optimization hoc Effects model performance and explainability **Explainability** 







2 [2 splits]



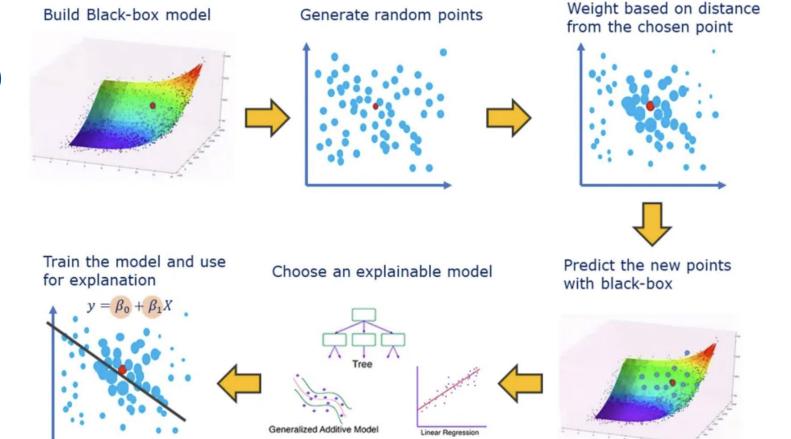
Number of weights, biases, & connections.

#### LIME is a post-hoc method for black-box models.

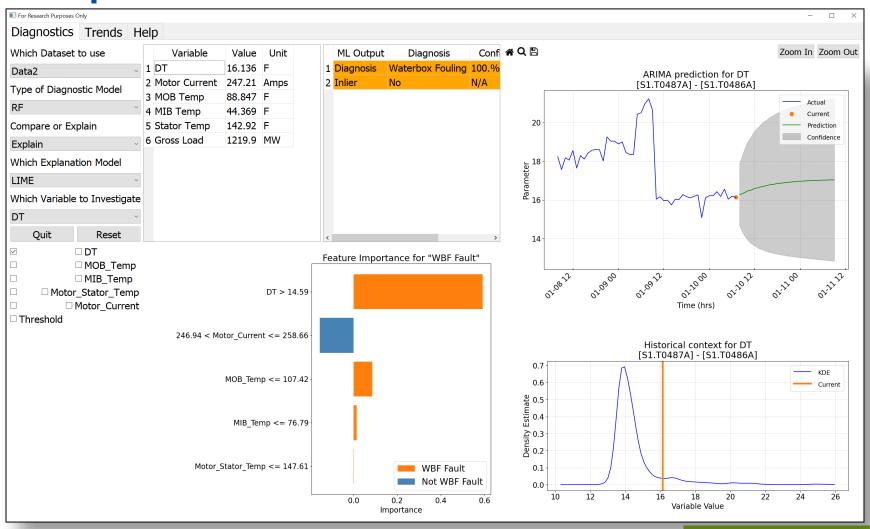
 Local interpretable modelagnostic explanations (LIME) can be used for any model.

• LIME is only valid locally.

 SHAP (Shapley Additive Explanations) are another common post-hoc method used to increase explainability.



## Model confidence, prognostics, explainability, and historical context all provide evidence for the conclusion.



#### Adding context to the data can further improve understanding.

