

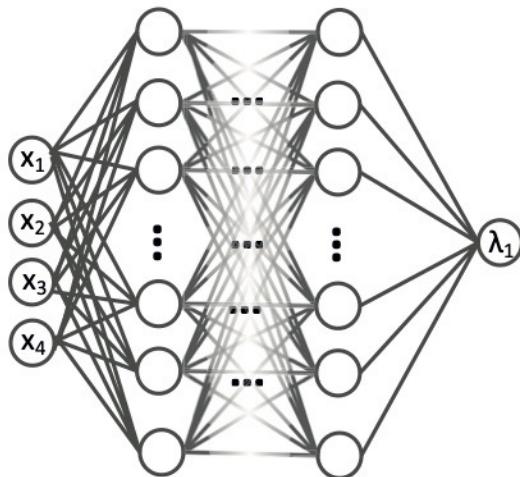


Machine Learning for Converting Black-Box Models to Interpretable Functions

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Motivation and Goal



Neural Network Representation of a Function λ

$$\lambda = P(x) = 4x_1^2x_3 - 3x_2 + x_4^3 - 5$$

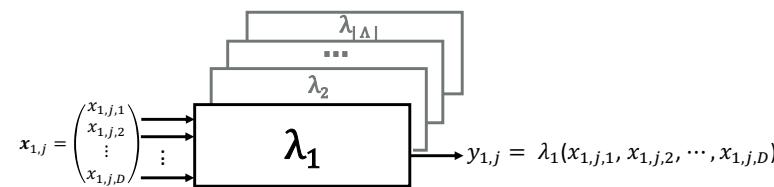
Polynomial Representation of a Function λ

- **Goal:** Find a mapping from the model internals (e.g. the weights and biases of a neural network) to semantically well defined and human-understandable domain

General Framework

1. Generate functions $\lambda_{1,2,\dots,|\Lambda|} \in \Lambda$ and datasets

$$\{\mathcal{D}_{\lambda_i} = \{(x_{ij}, y_{ij})\}_{j=1}^N\}_{i=1}^{|\Lambda|}$$



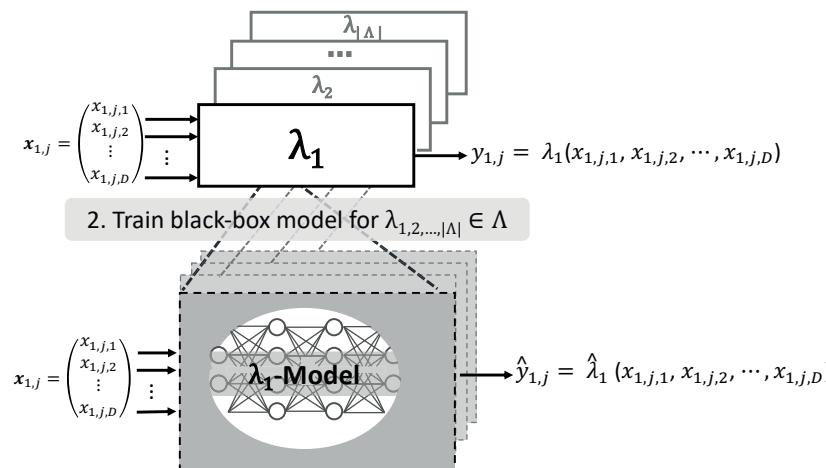
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3. Generate training data for I -Model

$$x_{1,j} = \begin{pmatrix} x_{1,j,1} \\ x_{1,j,2} \\ \vdots \\ x_{1,j,D} \end{pmatrix}$$



General Framework

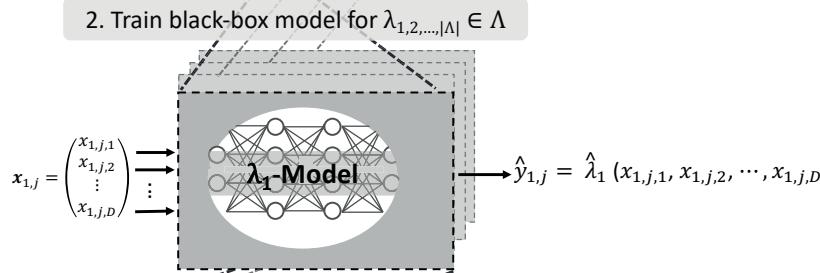
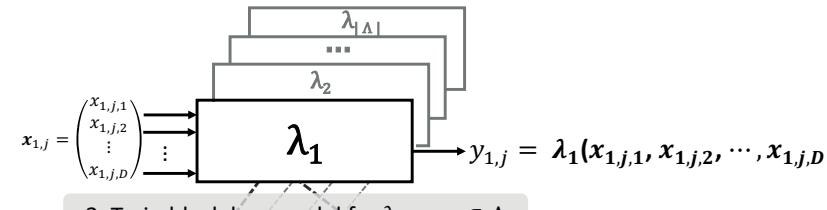
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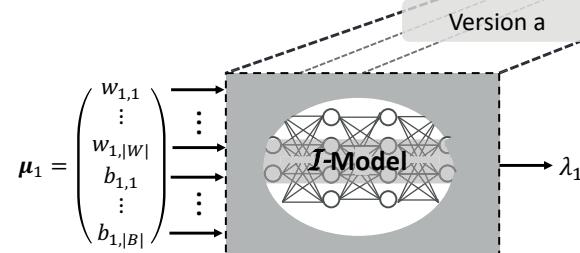
3. Generate training data for I -Model

a) $\mathcal{D}_{I_a} = \{(\mu_i, \lambda_i)\}_{i=1}^{|\Lambda|}$

b) $\mathcal{D}_{I_b} = \{(\mu_i, \lambda_i)\}_{i=1}^{|\Lambda|}$



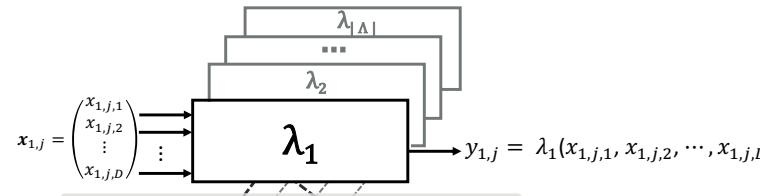
4. Train I -Model



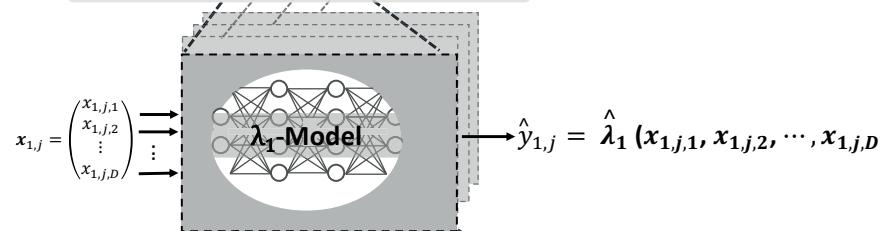
General Framework

1. Generate functions $\lambda_{1,2,\dots,|\Lambda|} \in \Lambda$ and datasets

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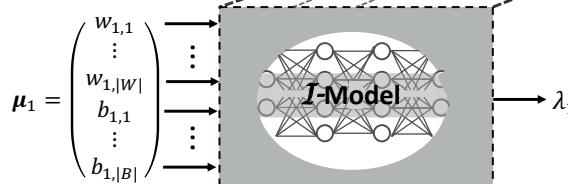
2. Train black-box model for $\lambda_{1,2,\dots,|\Lambda|} \in \Lambda$



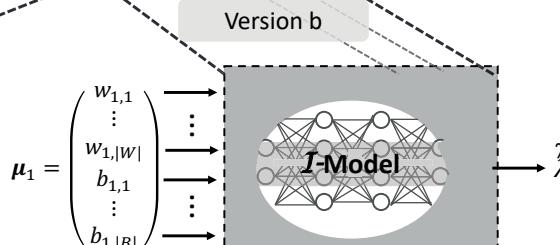
3. Generate training data for I -Model

$$a) \quad \mathcal{D}_{I_a} = \{(\mu_i, \lambda_i)\}_{i=1}^{|\Lambda|}$$

$$b) \quad \mathcal{D}_{I_b} = \{(\mu_i, \tilde{\lambda}_i)\}_{i=1}^{|\Lambda|}$$



4. Train I -Model



Conclusion

- Using a trained Interpretation-Model we are able to interpret arbitrary black-box models as terms of the previously estimated algebra
- To achieve good results, we need to make two assumptions:
 1. It is possible to approximate the model we want to interpret accurately using the estimated algebra
 2. We have found an appropriate representation for functions from this algebra