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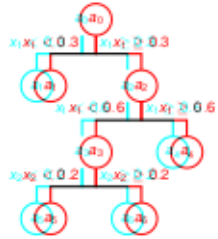

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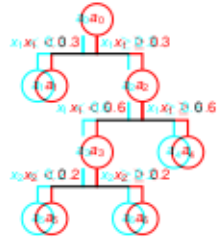

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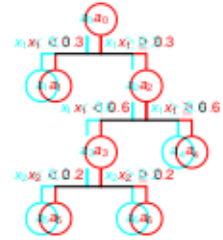

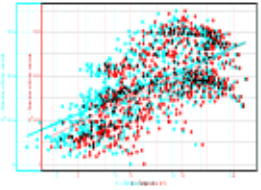

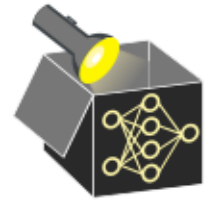

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- Many people are familiar with simple interpretable models
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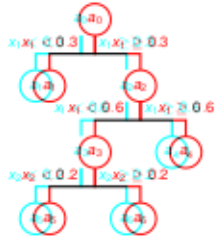

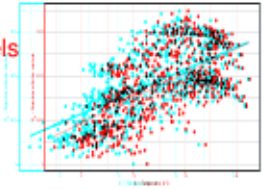

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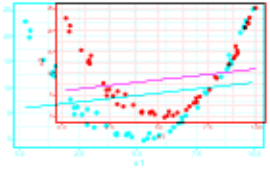

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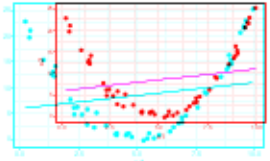

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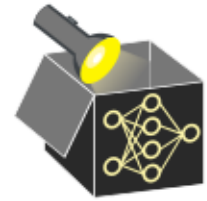

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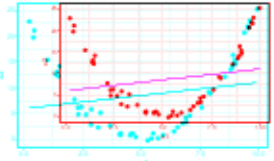

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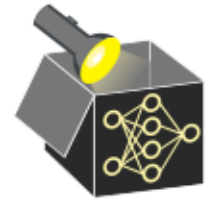

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- Often there is a trade-off between interpretability and model performance