

offer multiple standardization parameters, but the number of trees in the forest is highlighted here

The SVR model exhibited accuracy rates of 59%, whereas the XGBoost model exhibited accuracy rates of 87%. Therefore, according to the results, the SVR model can be considered relatively weak in comparison to the XGBoost regression model.

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