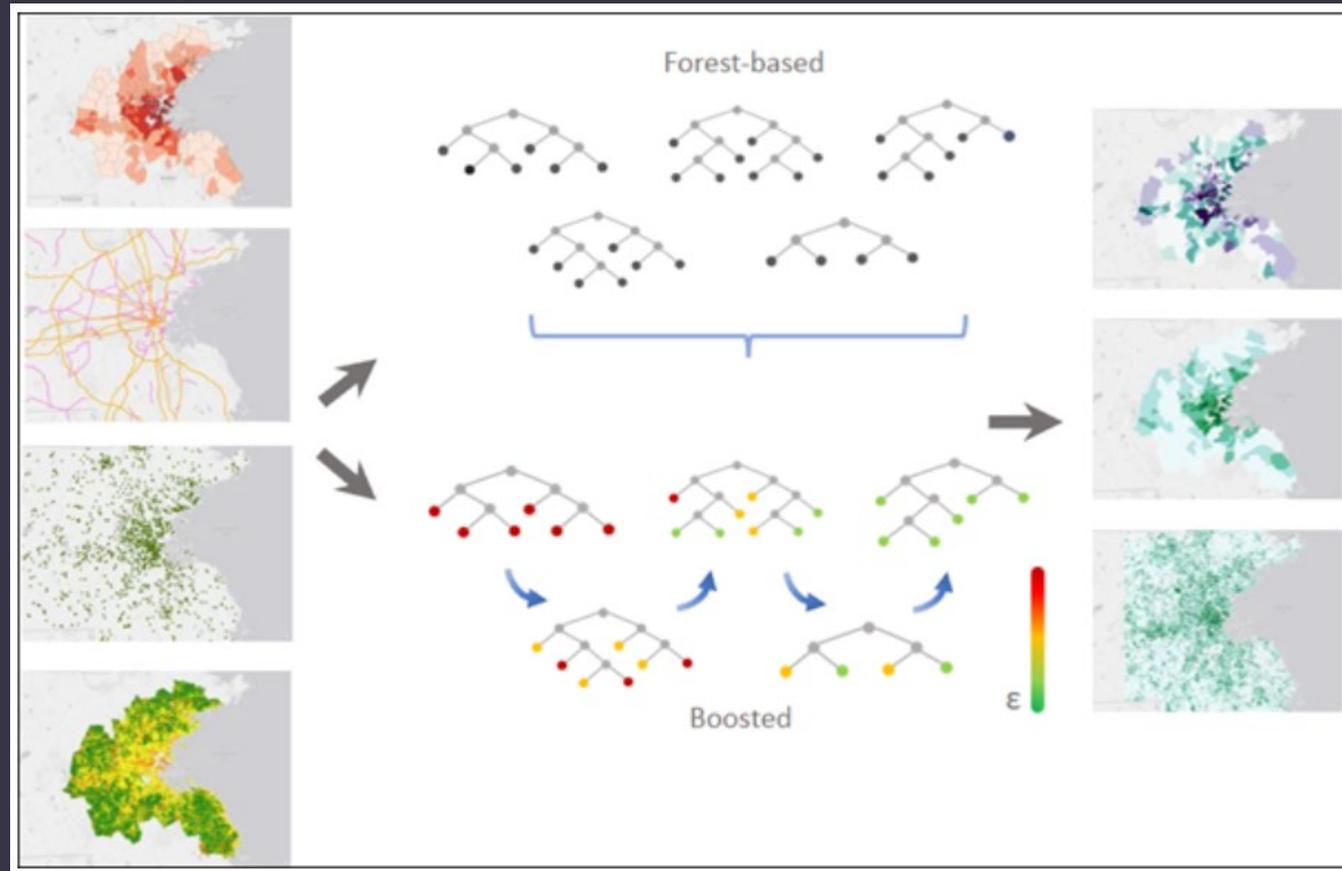


MODELING SPATIAL RELATIONSHIPS

Forest-based and Boosted Classification and Regression

RANDOM FOREST



FOREST BASED & GRADIENT BOOSTED MODELS

Forest based

- Many independent decision trees
- Each decision tree is created from a random subset of training data and explanatory variables
- Each tree generates its own predictions
- Final prediction is based on an average of all decision trees in the entire forest

Gradient boosted

- Creates a series of sequential decision trees
- Each tree is built to minimize the error (bias) of the previous tree
- Combines weak learners to become a strong prediction model

HOW TO TRAIN A MODEL

- Prediction Type
 - Train only
 - Predict to features
 - Predict to raster
- Explanatory Training Variables
- Explanatory Training Distance Features
- Explanatory Training Rasters

The screenshot shows the 'Parameters' tab of a geoprocessing tool. The tool name is 'Forest-based and Boosted Classification and Regression'. The 'Prediction Type' is set to 'Train only' and the 'Model Type' is 'Forest-based'. There are three required fields: 'Input Training Features', 'Variable to Predict', and 'Output Trained Model File', each with a folder icon. The 'Explanatory Training Variables' section has a 'Variable' field with a gear icon and a 'Categorical' checkbox. The 'Explanatory Training Distance Features' field has a dropdown arrow and a folder icon. The 'Explanatory Training Rasters' field has a folder icon and a 'Categorical' checkbox. At the bottom, there are three expandable sections: 'Additional Outputs', 'Advanced Model Options', and 'Validation Options'.

Geoprocessing

Forest-based and Boosted Classification and Regression

Parameters Environments

Prediction Type
Train only

Model Type
Forest-based

* Input Training Features

* Variable to Predict

Treat Variable as Categorical

Explanatory Training Variables

Variable Categorical

Explanatory Training Distance Features

Explanatory Training Rasters Categorical

Output Trained Model File

> Additional Outputs

> Advanced Model Options

> Validation Options

VALIDATION AND TRAINING DATA DIAGNOSTICS

- R-squared – proportion of variance in the dependent variable that can be explained by the independent variable
- Goodness of fit
- 0.0 to 1.0 (higher the better)
- While training, the tool reserves 10% of data for validation – it only uses 90% of the training data and uses validation data to determine how well the model performed preventing Overfitting.

Training Data: Regression Diagnostics

R-Squared	0.844
Mean Absolute Error (MAE)	33329.720
Mean Absolute Percentage Error (MAPE)	0.194
Root Mean Square Error (RMSE)	45848.739
p-value	0.000
Standard Error	0.003

*Predictions for the data used to train the model compared to the observed categories for those features

Validation Data: Regression Diagnostics

R-Squared	0.705
Mean Absolute Error (MAE)	44251.616
Mean Absolute Percentage Error (MAPE)	0.250
Root Mean Square Error (RMSE)	62644.450
p-value	0.000
Standard Error	0.010

*Predictions for the test data (excluded from model training) compared to the observed values for those test features

THINGS TO KEEP IN MIND...

- These tools can perform well within the range of explanatory variables used to train the model.
- Forest-based and boosted models do not extrapolate.
- Number of trees (100 by default) can be increased with the complexity of relationships between variables, size of dataset, and variable to predict.
- Tool runtime is sensitive to the number of variables used per tree.
- The tool may generate slightly different results each time due to the randomness introduced in the algorithm.

THINGS TO KEEP IN MIND...

- A “good” model is subjective and varies based on the data.
- Some industries may have a r-square value (threshold) to consider if the model is satisfactory for making a prediction.
- Refining a Model – includes only most important variables. The variables with low importance that did not contribute to the model performance were excluded.
- Increasing number of trees in a model could result in a better model.
- Model Out of Bag Errors – Determines how much the model performance has improved by increasing the number of trees in the model.