Random Forest with ranger using fold sampling

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Random Forest Regression using sample summaries runs faster, retains accuracy

Summarising or Folding the samples is a way of reducing the total number of samples to a manageable number in order to run prediction algorithms on modern day machines. The folded samples are then unfolded to predict the full dataset.

```
# Removes one feature at a time and uses it as the variable to be predicted (y variable)

# Total permutations:

permutation_test_number <- 100

# Run SVD decomposition of samples to a reduced sample space
source("./decompose_sample_space.R")

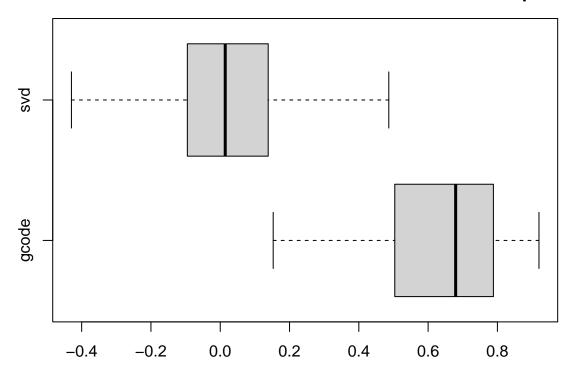
# Run gcode encoding of samples to a reduced sample space
source("./encode_sample_space.R")</pre>
```

A way to fold the total number of samples while retaining the original sample structure is done via Generative Encoding (gcode): $\frac{\text{https://github.com/AskExplain/gcode/tree/alpha_test_v2022.1} \\$

Boxplots of Gini importance Correlation for the features after summarisation compared to using all samples

In essence, the Gini Importance is calculated for the features using all samples, and then correlated with the corresponding Importance values when using the summary samples from either SVD or gcode. The output is a correlation R^2 value for each run, giving a distribution for multiple runs on different permutations of the training set.

Gini importance
Pearson Correlation of summaries relative to all samples

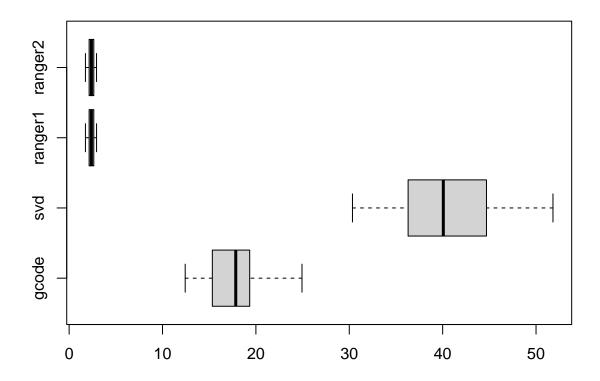


Boxplots of mean absolute error and runtime are plotted for every unique run of random forest regression via ranger package.

```
# Mean Absolute Error
boxplot(data.frame(gcode = total_mae.gcode,
```

Of great importance, the runtime does not include the running of the SVD or gcode algorithms.

Mean Absolute Error distribution



Runtime distribution

