

Multilayer perceptron with RSNNS package using fold sampling

David Banh - AskExplain

04/01/2022

Multilayer perceptron 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 <- 10

# 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")
```

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

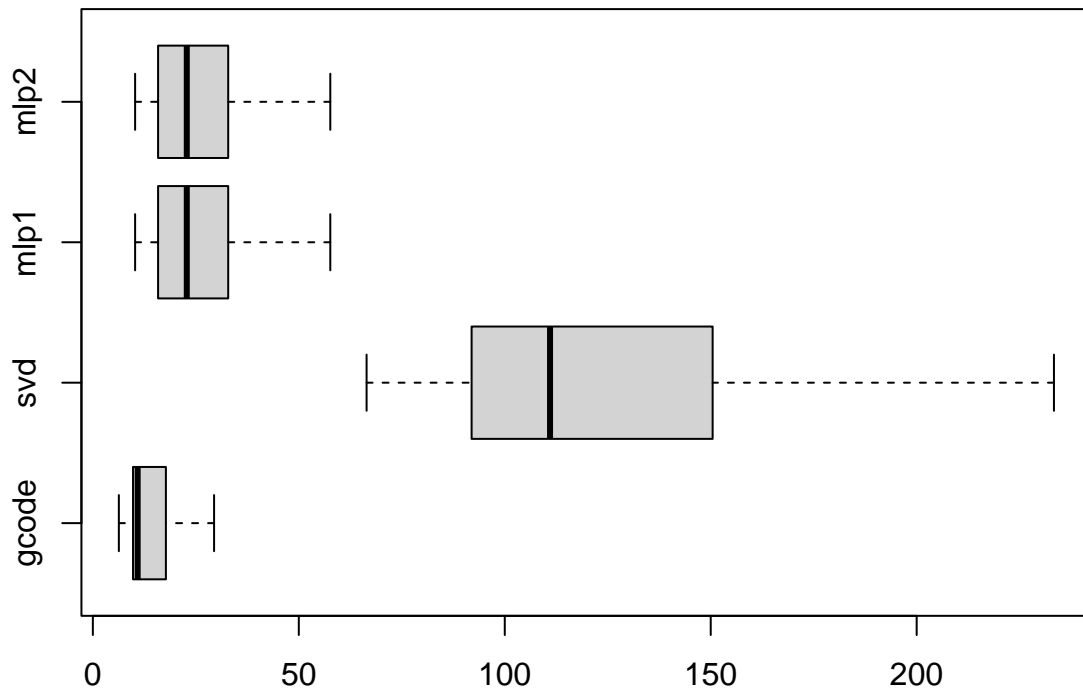
Boxplots of mean absolute error and runtime are plotted for every unique run of the neural network regression via RSNNS package.

```
# Mean Absolute Error

boxplot(data.frame(gcode = (total_mae.gcode),
                  svd = (total_mae.svd),
                  mlp1 = (total_mae.lreg.1),
                  mlp2 = (total_mae.lreg.2)),
        outline = F,
        horizontal = T,
        main = "Mean Absolute Error distribution")
```

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

Mean Absolute Error distribution



```
# Runtime
```

```

boxplot(data.frame(gcode = total_time.gcode,
                  svd = total_time.svd,
                  mlp1 = total_time.lreg.1,
                  mlp2 = total_time.lreg.2),
        outline = F,
        horizontal = T,
        main = "Runtime distribution")

```

