Regel-gebaseerde regressie Capita Selecta computerwetenschappen Artificiële intelligentie (|H05N0a|)

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Overzicht

- optimalisatiecriterium Gecode: mean squared error
- regressiemodel: beslissingsboom + pruning

Beslissingsboom

end

Algorithm 1: buildRegressionModel(data, itemsets, items)

```
if nb instances of data ≤ MAX_NB_LEAF then
    prediction ← average (OR mean) of class values of instances
    add items to itemsets (as last rule)
else
    items2 ← find best pattern with rimcp on data
    items2 ← items2 ∪ items
    dataC ← instances of data that cover items in items2
    dataNC ← instances of data that don't cover items in items2
    buildRegressionModel(dataC, itemsets, items2)
    buildRegressionModel(dataNC, itemsets, items)
```

Pruning

Algorithm 2: pruneRules(data, predOr, itemsets)

```
newPredOr ← predOr
repeat
  itemsets ← pruneRule(data, newPredOr, items)
  apply rules of itemsets on newPredOr
until size new ruleset == size old ruleset
return itemsets
```

Algorithm 3: pruneRule(data, predOr, itemsets)

```
\begin{array}{l} bestRulesSoFar \leftarrow itemsets \\ tempData \leftarrow predOr \\ bestError \leftarrow error of predOr on data \\ \textbf{for } each rule in itemsets \ \textbf{do} \\ tempRules \leftarrow itemsets \setminus rule \\ apply rules of tempRules on tempData \\ error \leftarrow error of tempData on data \\ \textbf{if } (error - RULECOST) < bestError \ \textbf{then} \\ bestError = (error - RULECOST) \\ bestRulesSoFar \leftarrow tempRules \\ \textbf{end} \\ \end{array}
```

return bestRulesSoFar

| prune | RULECOST | MAX_NB_LEAF | nb of rules | error trainingset | error testset |
|-------|----------|-------------|-------------|-------------------|---------------|
| false | / | 3 | 33 | 67.33 | 158.77 |
| true | 5 | 3 | 13 | 87.90 | 158.47 |
| false | / | 5 | 20 | 85.92 | 143.73 |
| true | / 5 | 5 | 12 | 99.04 | 142.38 |
| false | / | 7 | 15 | 92.35 | 141.78 |
| true | 5 | 7 7 | 10 | 106.33 | 139.90 |
| false | / | 10 | 11 | 102.05 | 139.84 |
| true | 5 | 10 | 9 | 109.06 | 137.96 |
| true | 0 | 3 | 33 | 67.33 | 158.77 |
| true | 0 | 5 | 20 | 85.92 | 143.73 |
| true | 0 | 7 | 15 | 92.35 | 141.78 |
| true | 0 | 10 | 11 | 102.05 | 139.84 |
| true | 3 | 3 | 15 | 80.13 | 158.47 |
| true | 3 | 5 7 | 14 | 92.03 | 143.73 |
| true | 3 | 7 | 12 | 99.35 | 141.78 |
| true | 3 | 10 | 11 | 102.05 | 139.84 |
| true | 10 | 3 | 12 | 94.9 | 158.47 |
| true | 10 | | 8 | 124.32 | 142.38 |
| true | 10 | 5 7 | 8 | 118.53 | 143.39 |
| true | 10 | 10 | 8 | 114.29 | 137.96 |

Tabel: Voorspelling met gemiddelde. Invloed van de parameters op de training en test set error.

| prune | RULECOST | MAX_NB_LEAF | nb of rules | error trainingset | error testset |
|-------|----------|-------------|-------------|-------------------|---------------|
| false | / | 3 | 33 | 54.20 | 127.00 |
| true | 5 | 3 | 11 | 89.30 | 128.20 |
| false | / | 5 | 20 | 67.60 | 125.55 |
| true | 5 | 5 | 11 | 89.65 | 126.90 |
| false | / | 10 | 11 | 91.10 | 118.15 |
| true | 5 | 10 | 10 | 99.04 | 126.40 |

Tabel: Voorspelling met mediaan. Invloed van de parameters op de training en test set error.