

# Implementing Tuple Variables in Gecode

Bachelor Thesis  
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# Constraint Programming

- Declarative style programming
- The programmer states what must hold, without stating how it is achieved
- A constraint programming variable has more in common with mathematical variables than regular variables used in programming

# Constraint Programming

while not done:

    propagators prune as much as possible

    a brancher assigns a value to a variable

# Constraint Programming

Sudoku is a good example of a problem suitable for constraint programming.

# Constraint Programming

```
IntVar s[9][9] = {1, ... ,9}
```

```
for x=1 to 9:
```

```
    distinct(s[x][1], s[x][2], ... , s[x][9])
```

```
for y=1 to 9:
```

```
    distinct(s[1][y], s[2][y], ... , s[9][y])
```

```
for x=1 to 3:
```

```
    for y=1 to 3:
```

```
        distinct( s[3x-2][3y-2], s[3x-1][3y-2], s[3x][3y-2],  
                  s[3x-2][3y-1], s[3x-1][3y-1], s[3x][3y-1],  
                  s[3x-2][3y], s[3x-1][3y], s[3x][3y])
```

# What data types exist in Gecode?

- Boolean
- Integer
- Integer set
- Float

# What about Tuples?

# What about Tuples?

Consider the following example:

```
Intvar a,b = {1, ..., 1000}
```

```
rel(a ≠ b)
```



# What about Tuples?

- Without tuples, it is impossible to prune a combination of values.
- Tuples move work from branchers to propagators.

# Pair Variables

- We restrict tuples to 2-tuples and call them pairs
- Two variants are created
  - Exact domain representation
  - Approximate representation

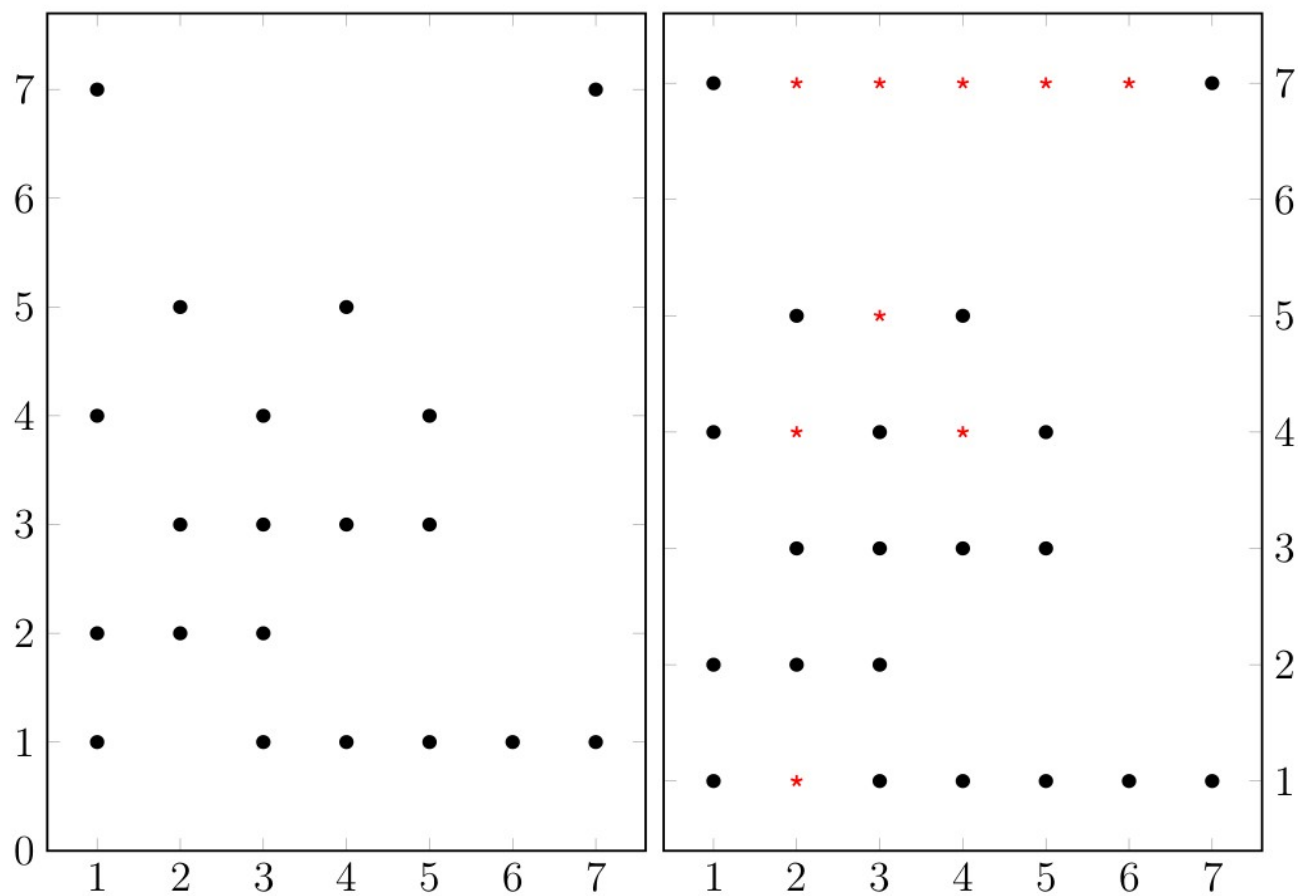
# Exact Pair

- Stores a long list of ALL combinations
- Requires more memory
- Some operations are slower
- Can be pruned more efficiently

# Approximate Pair

- Does not store all possible combinations
- Requires less memory
- Some operations are faster
- Can not be pruned as efficiently as the exact representation

# Comparison



# The cDFA constraint

- A cDFA is a DFA with individual costs for each transition
- A cDFA does not only tell if a string is accepted or not, but also the cost

# The cDFA constraint

$\text{cDFA}(ps, pc, qs, qc, x, S, C)$

- $qs$ : State before the transition
- $qc$ : Cost before the transition
- $x$ : Next symbol in the string
- $S$ : State function
- $C$ : Cost function
- $ps = S(qs, x)$
- $pc = qc + C(qs, x)$

# The cDFA constraint

$\text{cDFA}(P, Q, x, S, C)$

- $Q$ : State and cost before the transition
- $x$ : Next symbol in the string
- $S$ : State function
- $C$ : Cost function
- $P = \langle S(Q.s, x), Q.c + C(Q.s, x) \rangle$



# Comparing performance

```
PairVar P[n+1]
```

```
IntVar x[n]
```

```
for i=1 to n:
```

```
    cDFA(P[n+1], P[n], x[n], S, C)
```

```
IntVar s[n+1], c[n+1], x[n+1]
```

```
for i=1 to n:
```

```
    cDFA(s[n+1], c[n+1], s[n], c[n],  
        x[n], S, C)
```

# Comparing performance

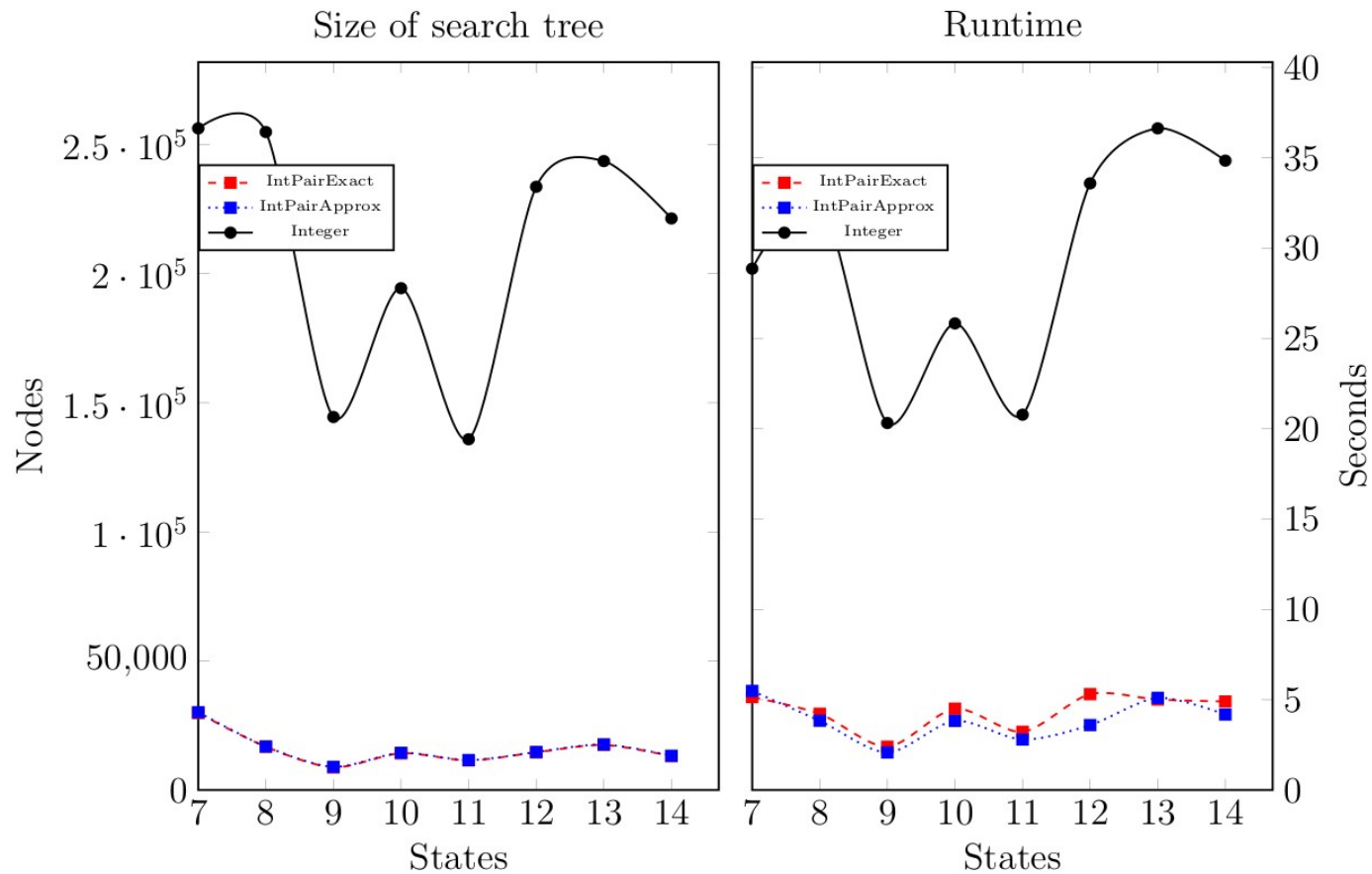
Performance is measured in two ways:

- Size of search tree
- Total runtime

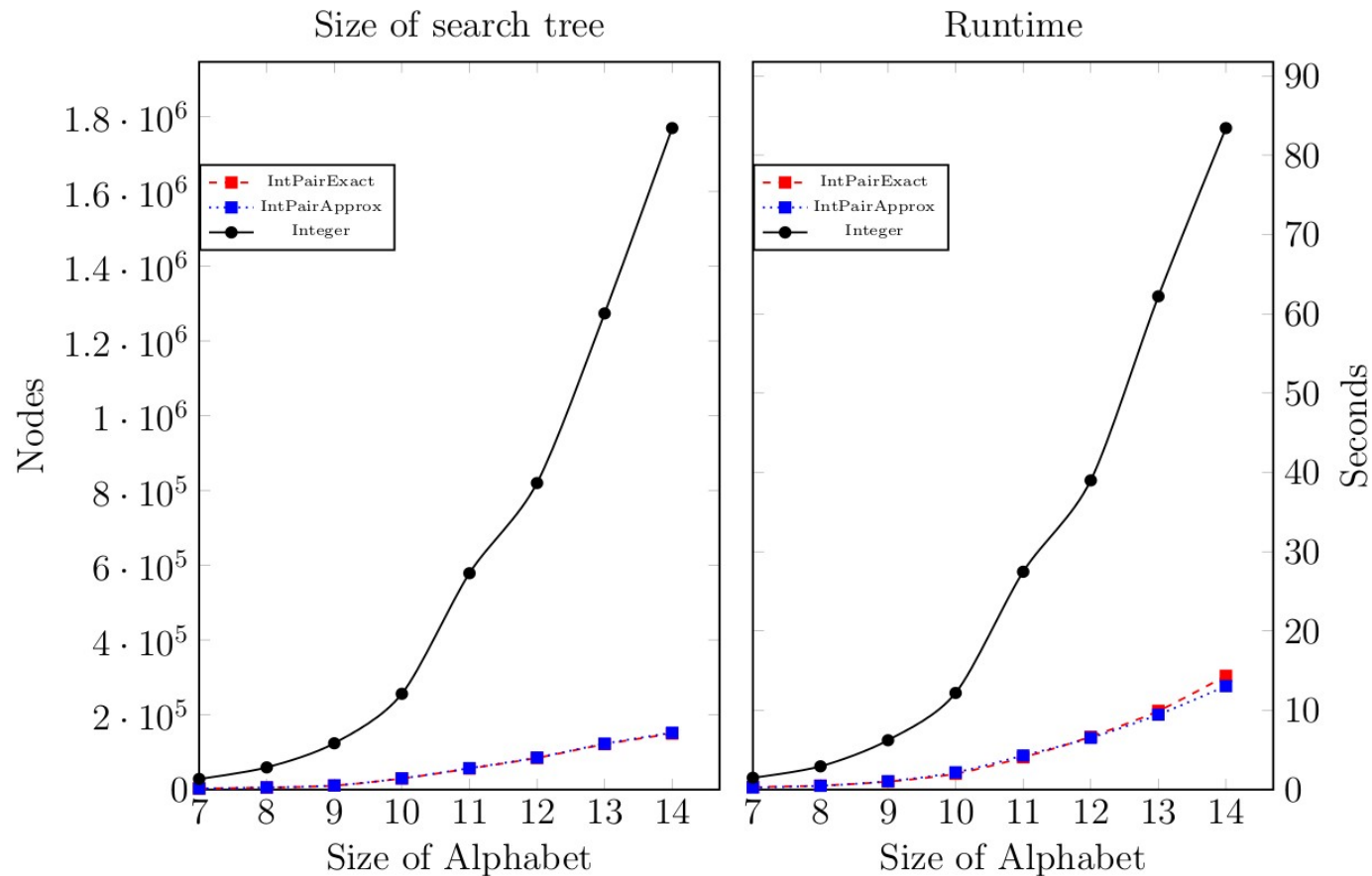
Performance is compared by varying four parameters:

- Number of states
- Size of alphabet
- Cost per transition
- Length of string

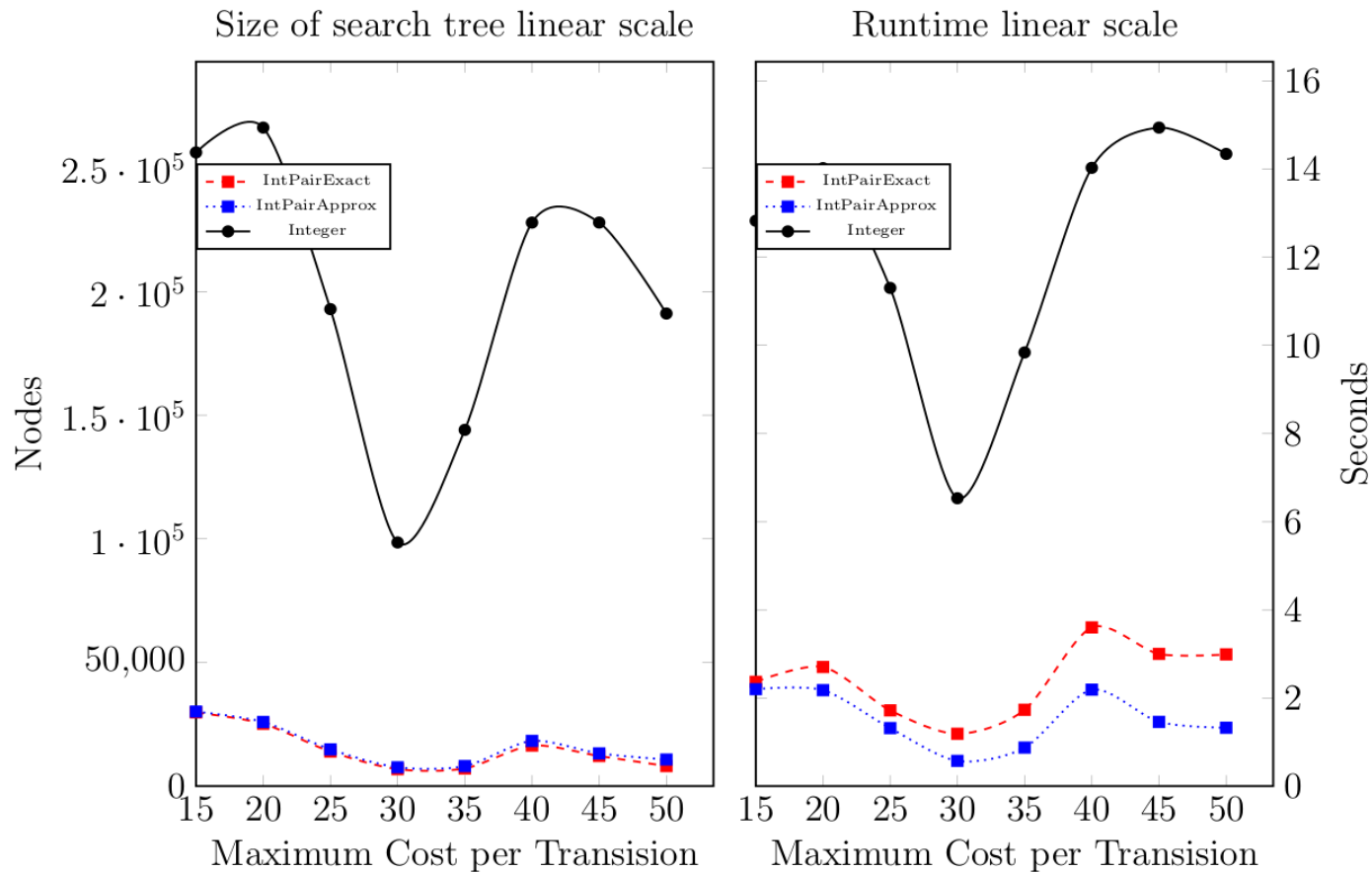
# Varying number of states



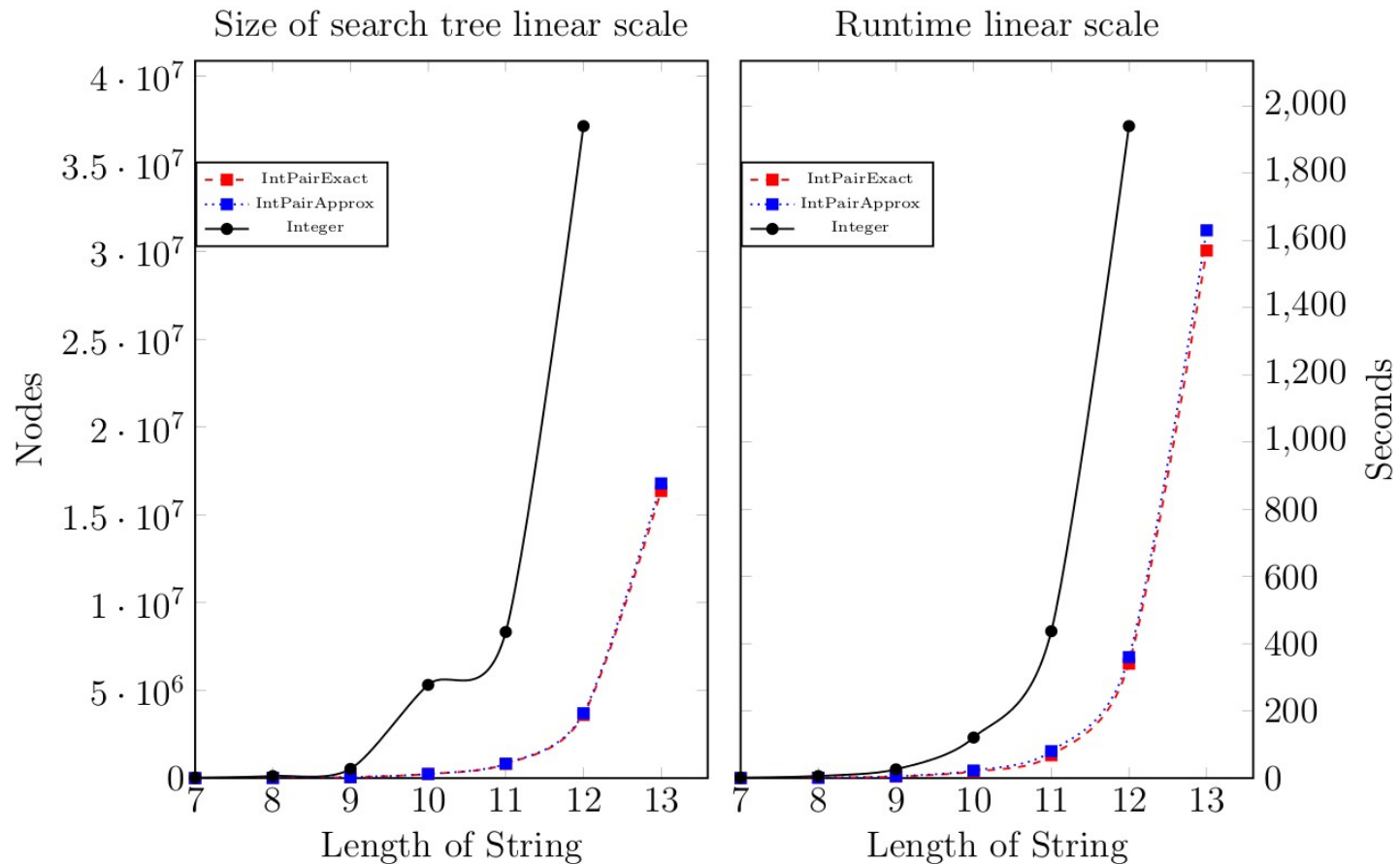
# Varying size of alphabet



# Varying cost per transition



# Varying length of string



# Possible Problems

The propagator for regular integer variables may be optimised further. However, much more work have been spent on optimising this.

# Conclusions

Tuples seem very promising. In these tests, they perform far better than regular integer variables. This is true for both the size of search tree and the total execution time.