Using Model Checking with Symbolic Execution for the Verification of Data-Dependent Properties of MPI-Based Parallel Scientific Software

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Problem

- It is hard to create “correct” parallel programs
  - Concurrency adds complexity and introduces problems such as deadlock
  - Non-determinacy makes testing even less effective
- Model checking techniques have been applied to concurrent systems
  - But focus on patterns of communication instead of correctness of the computation
  - Limited experience with MPI-based programs
Objective

• Verification of MPI-based programs using model checking
  – Freedom from Deadlock
  – Computational correctness
Approach

• Use a model checker to explore all possible executions

• Deadlock detection
  – Modeling MPI functions
  – Abstracting away unnecessary data

• Computational correctness
  – Extend the model checker to create symbolic representations of the executions
  – Compare the sequential program’s symbolic representation of the result to the parallel program’s representation
Outline

• Background on MPI
• Verification for freedom from deadlock
• Verification using symbolic execution
• Experimental results
Overview of MPI

• Origin: result of work of the Committee formed at the Supercomputing ’92 conference
• Significance: widely used in scientific computation
• Communication: message passing in a distributed-memory environment
• Types of operations supported:
  – Blocking/non-blocking
  – Collective
Example MPI Program

Multiplication of two 3X3 matrices A and B with two processes

Proc 0
A
B

Proc 1
A
B

Proc 0
A
\[
\begin{bmatrix}
X & 0 & X \\
0 & X & 0 \\
X & 0 & X \\
\end{bmatrix}
\]

Proc 1
A
\[
\begin{bmatrix}
0 & X & 0 \\
X & 0 & X \\
0 & X & 0 \\
\end{bmatrix}
\]
Example MPI Program

Proc 0
A
B
C =

Proc 1
A
B
C =

MPI_Reduce

Proc 0
A
B
C
D =

Proc 1
A
B
C
Example MPI Program: Code

double *A, *B;  //two initial matrices
double *C;     //intermediate result
double *D;     //final result, D = AB

/* Broadcast entries of A and B */
MPI_Bcast(A, nElts, MPI_DOUBLE, 0, MPI_COMM_WORLD);
MPI_Bcast(B, nElts, MPI_DOUBLE, 0, MPI_COMM_WORLD);

/* Read matrix dimensions from file*/
if(rank == 0){
    fscanf(inFile, "%d\n", &n);
    nElts = n*n;
}

/* Broadcast the dimension and the total number of
   elements in each matrix*/
MPI_Bcast(&nElts, 1, MPI_INT, 0, MPI_COMM_WORLD);
MPI_Bcast(&n, 1, MPI_INT, 0, MPI_COMM_WORLD);

/* Allocate memory for A, B, and C*/
A, B, C = (double *) malloc(nElts * sizeof(double));

/* Root process allocates memory for D and reads
   entries of A and B from file */
if (rank==0){
    D = (double *) malloc(nElts * sizeof(double));
    for(i=0; i<nElts; i++) {
        if (i%n==0) fscanf(inFile, "\n");
        fscanf(inFile, "%lf", &A, B[i]);
    }
}

/* Computation */
for (i = 0; i<n; i++)
    for (j=0; j<n; j++) {
        C[(i*n)+j]=0;
        if (((i*n)+j)%size==rank) for (k = 0; k < n; k++) C[(i*n)+j] +=
            (double)A[(i*n)+k] * (double)B[(k*n)+j];
    }

/* Reduce entries in C to the corresponding entries D on
   the root process*/
for (i = 0; i<nElts; i++)
    MPI_Reduce(&C[i], &D[i], 1, MPI_DOUBLE, MPI_MAX, 0,
               MPI_COMM_WORLD);

/* Broadcast the dimension and the total number of
   elements in each matrix*/
MPI_Bcast(&nElts, 1, MPI_INT, 0, MPI_COMM_WORLD);
MPI_Bcast(&n, 1, MPI_INT, 0, MPI_COMM_WORLD);

/* Allocate memory for A, B, and C*/
A, B, C = (double *) malloc(nElts * sizeof(double));

/* Root process allocates memory for D and reads
   entries of A and B from file */
if (rank==0){
    D = (double *) malloc(nElts * sizeof(double));
    for(i=0; i<nElts; i++) {
        if (i%n==0) fscanf(inFile, "\n");
        fscanf(inFile, "%lf", &A, B[i]);
    }
}
Verification Process Overview

• Model checking
• Modeling the program
  – MPI functions
  – Deadlock free
  – Computation correctness
• Carrying out the verification
Typical Model Checking Architecture

- **Program (C, MPI)**
- **Program Model (PROMELA)**
- **Model Checker (SPIN)**

- **Property**
  - Usually represented in temporal logic or sometimes as an automaton
  - Builds a graph representing all possible states of the system
  - State explosion problem: number of states is exponential in the number of concurrent processes
  - Hence necessary to create an abstracted version of the program to eliminate unnecessary details wrt the property being proved

- **abstraction**
- **property verified**
- **counter example (execution trace) produced**
Example Code

• **Matrix Multiplication** -
  *written by my classmates and myself as one of the assignments for the Programming Languages Course in Computer Science (CS331) at the University of Alaska, Anchorage*

• **Gauss-Jordan Elimination** –
  *written at the University of Massachusetts, Amherst under the guidance of Dr. Stephen Siegel*

Testing was performed on both of these implementations and based on the results the code was assumed to be correct.
Program Model: MPI

Key Abstractions:

• Processes and communication channels
• Collective MPI functions
• Messages
Processes and Channels in PROMELA

- Definition of Processes:
  
  ```
  active proc Root{...}
  active proc [numprocs - 1] NonRoot{...}
  active proc Coordinator{...}
  ```

- Definition of Channels:
  
  ```
  /*Data channels from root to non-root nodes*/
  chan chan_from_root[nprocs]
  /*Data Channels from non-root nodes to root */
  chan chan_to_root[nprocs]
  /*Collective communication channels from nodes to Coordinator */
  chan chan_c[nprocs]
  ```
Modeling MPI Functions

Matrix multiplication example utilizes two MPI functions, both of which are collective:

- **MPI_Bcast**(void *buffer, int count, MPI_Datatype datatype, int root, MPI_Comm comm)
  \(\text{Broadcasts a message from the process with rank root to all processes of the group}\)

- **MPI_Reduce**(void *sendbuf, void *recvbuf, int count, MPI_Datatype datatype, MPI_Op op, int root, MPI_Comm comm)
  \(\text{Combines the elements provided in the input buffer of each process in the group, using the operation op, and returns the combined value in the output buffer of the process with rank root}\)

Based on the description in Siegel and Avrunin, introduce another process for coordinating collective operations…
Modeling MPI Functions: Coordinator Process

The Coordinator process is used to model the collective MPI functions MPI_Bcast and MPI_Reduce:

**MPI_Bcast**

- Proc 0
- Proc 1
- …
- Proc n

…wait for all processes to receive data from root…

- Proc 0
- Proc 1
- …
- Proc n

DONE

**MPI_Reduce**

- Proc 0
- Proc 1
- …
- Proc n

…wait for the root process to collect data from all procs…

- Proc 0
- Proc 1
- …
- Proc n

DONE
Coordinator Process: PROMELA

Code

active proctype Coordinator() {
  /* Receive initial messages from all processes and confirm that all nodes agree on the type of communication in progress*/
  do
    :: chan_c[0]?im;
    i = 1;
    do
      :: i < nprocs -> chan_c[i]?cm; assert(im == cm); i++
      :: else -> i = 0; break
    od;
    if
      :: im == BCAST ->
        /* Interpret a broadcast message, receive a DONE message from every node */
        do
          :: i < nprocs -> chan_c[i]?cm; assert(cm == DONE); i++
          :: else -> i = 0; break
        od
      :: im == REDUCE ->
        /* Interpret a reduce message, receive a DONE message from the root process */
        ...
      fi
    od
}
Messages

Messages between the coordinator and processes participating in the computation are abstracted in the following manner:

\[ \text{mtype} = \{ \text{BCAST, REDUCE, DONE} \}, \]

where

- **BCAST** is a message sent to the Coordinator process by every process participating in the computation at the beginning of all the MPI_Bcast collective routines
- **REDUCE** is sent to the Coordinator process by every process participating in the computation at the beginning of all the MPI_Reduce collective routines
- **DONE** is sent to the Coordinator process by processes participating in the computation upon completion of every collective operation as expected by the Coordinator process
Modeling to Check Freedom From Deadlock

• Abstractions to remove details about data
  
mtype = \{BCAST, REDUCE, DONE, \textbf{DATA}\}

Example:
  
  MPI\_Bcast (0, \textbf{DATA});
  
  - process with rank 0 broadcasting an abstracted data unit to all
    process participating in the computation

• Synchronization issues
  
  channel!message
  if
    :: 1 \rightarrow \text{empty}(channel)
    :: 1 \rightarrow \text{skip}
  fi
Check for Freedom from Deadlock

• Apply SPIN
• No need for special property definition
• Scalability
  – raw
  – utilizing optimizations of SPIN results in extended capabilities
Modeling Computation

- MPI Communication – the same as in the deadlock model
- Add support for symbolic expressions in PROMELA
  - Symbolic expressions are stored in byte arrays in prefix notation

```c
typedef Expn{
  byte length;
  byte array[maxExpnSize]
}
```

Example:

\[
\begin{array}{ccc}
a_0 & a_1 & a_2 \\
a_3 & a_4 & a_5 \\
a_6 & a_7 & a_8 \\
\end{array}
\]

\[
\begin{array}{ccc}
b_9 & b_{10} & b_{11} \\
b_{12} & b_{13} & b_{14} \\
b_{15} & b_{16} & b_{17} \\
\end{array}
\]

\[
a_2 \cdot a_3 + a_4 \cdot a_5 \text{ in infix notation is equivalent to } +\cdot a_2 a_3 \cdot a_4 a_5 \text{ in prefix notation}
\]

and to \[
\begin{array}{cccccc}
+ & * & 2 & 3 & * & 4 & 5
\end{array}
\]

in the model

* Actually, integer constants represent operators, e.g. \#define 255 PLUS
Modeling Computation During Verification

- Add functions to manipulate symbolic expressions
  
  inline addTo(a, b){…} results in $a = a + b$
  inline multBy(a, b){…} results in $a = a \times b$

Example:

$$
\begin{align*}
\text{a} & = 2 \times 3 \\
\text{b} & = 4 \times 5 \\
\text{a+b} & = (2 \times 3) + (4 \times 5)
\end{align*}
$$
Validating the Results

Key Idea:
1. Implement both the sequential and parallel versions of the algorithm
2. Apply SPIN and compare the symbolic expressions produced
3. Conclude that the result of the parallel computations is correct if it matches the output of the sequential code.

Matrix multiplication example:
1. Symbolic computation is performed in parallel, generating a matrix of expressions on the root process
2. The root process does the symbolic computations sequentially
3. The root process loops through the two resultant structures checking that the two are the same via a set of assertions
More Interesting Example: Gauss-Jordan Elimination

Common application finding a matrix inverse

\[
\begin{bmatrix}
  a_{11} & \cdots & a_{1n} & 1 & 0 & \cdots & 0 \\
  a_{21} & \cdots & a_{2n} & 0 & 1 & \cdots & 0 \\
  \vdots & \ddots & \vdots & \vdots & \ddots & \ddots & \vdots \\
  a_{n1} & \cdots & a_{nn} & 0 & 0 & \cdots & 1
\end{bmatrix}
\]

where \(I\) is the identity matrix, to obtain a matrix of the form

\[
\begin{bmatrix}
  1 & 0 & \cdots & 0 & b_{11} & \cdots & b_{1n} \\
  0 & 1 & \cdots & 0 & b_{21} & \cdots & b_{2n} \\
  \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
  0 & 0 & \cdots & 1 & b_{n1} & \cdots & b_{nn}
\end{bmatrix}
\]

is then the inverse of \(A\) if such exists and \([A\ I]\) is in the reduced row-echelon form.
More Interesting Example: Gauss-Jordan Elimination

**Definition:** reduced row-echelon form

Properties of a matrix in a reduced row-echelon form:

1. Each row contains only zeros until the first non-zero element, which must be a 1
2. As the rows are followed from top to bottom, the first nonzero number occurs further to the right than in the previous row
3. The entries above and below the first 1 in each row must be all 0

Performing the Gauss-Jordan Elimination algorithm on any matrix results in the equivalent reduced row-echelon form of that matrix.
Gauss-Jordan Elimination: Example

Step 1. Locate the leftmost column (vertical line) that does not consist entirely of zeros

\[
\begin{bmatrix}
0 & 0 & -2 & 0 & 7 & 12 \\
2 & 4 & -10 & 6 & 12 & 28 \\
2 & 4 & -5 & 6 & -5 & -1 \\
\end{bmatrix}
\]

Leftmost nonzero column
Gauss-Jordan Elimination: Example

Step 2. Interchange the top row with another row, if necessary, so that the entry at the top column found in Step 1 is different from zero

\[
\begin{bmatrix}
0 & 0 & -2 & 0 & 7 & 12 \\
2 & 4 & -10 & 6 & 12 & 28 \\
2 & 4 & -5 & 6 & -5 & -1 \\
\end{bmatrix}
\]

Interchange the first and second rows
Gauss-Jordan Elimination: Example

Step 3. If the entry that is now at the top of the column found in Step 1 is \( a \), multiply the first row by \( \frac{1}{a} \) in order to introduce a leading 1.

\[
\begin{bmatrix}
1 & 2 & -50 & 6 & 62 & 28 \\
0 & 0 & -2 & 0 & 7 & 12 \\
2 & 4 & -5 & 6 & -5 & -1 \\
\end{bmatrix} \times \frac{1}{2}
\]
Gauss-Jordan Elimination: Example

Step 4. Add suitable multiples of the top row to the rows above and below so that all entries below leading 1 become zeros

\[
\begin{bmatrix}
1 & 2 & -5 & 3 & 6 & 14 \\
0 & 0 & -2 & 0 & 7 & 12 \\
0 & 0 & 55 & 0 & -17 & -29 \\
\end{bmatrix}
\]

2 times the first row was added to the third row
Gaussian-Jordan Elimination Algorithm

The above procedure is repeated for every row of the matrix to produce the reduced row-echelon form of the example matrix:

Initial matrix:

\[
\begin{bmatrix}
0 & 0 & -2 & 0 & 7 & 12 \\
2 & 4 & -10 & 6 & 12 & 28 \\
2 & 4 & -5 & 6 & -5 & -1
\end{bmatrix}
\]

Reduced Row-Echelon Form of the initial matrix:

\[
\begin{bmatrix}
1 & 2 & 0 & 3 & 0 & 7 \\
0 & 0 & 1 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 1
\end{bmatrix}
\]
Matrix Multiplication vs. Gauss-Jordan Elimination

Why is it harder to verify the correctness of the Gauss-Jordan Elimination?

- Need to introduce branching when conditions are functions of data
- The property is harder to express since there is no closed formula of the answer, again, consequence of data dependencies
Example of Symbolic Gauss-Jordan Elimination on a 2X2 matrix

A = \[
\begin{bmatrix}
  a_{11} & a_{12} \\
  a_{21} & a_{22}
\end{bmatrix}
\]

**Case 1:**
\[
\begin{align*}
  a_{11} &= 0 \\
  a_{21} &= 0 \\
  a_{12} &= 0 \\
  a_{22} &= 0
\end{align*}
\]

**Case 2:**
\[
\begin{align*}
  a_{11} &= 0 \\
  a_{21} &= 0 \\
  a_{12} &= 0 \\
  a_{22} &\neq 0
\end{align*}
\]

**Case 3:**
\[
\begin{align*}
  a_{11} &= 0 \\
  a_{21} &= 0 \\
  a_{12} &\neq 0 \\
  a_{22} &\text{free}
\end{align*}
\]

**Case 4:**
\[
\begin{align*}
  a_{11} &= 0 \\
  a_{21} &\neq 0 \\
  a_{12} &\text{free} \\
  a_{22} &\text{free}
\end{align*}
\]
Dealing with Data Dependency

Exploring Both Branches:

PROMELA Code:

```plaintext
if
:: 1 -> /* Assume (a_{11} == 0) */
/* Add expression (a_{11} == 0) to the path conditions table */
...
:: 1 -> /* Assume (a_{11} != 0) */
/* Add expression (a_{11} != 0) to the path conditions table */
...
fi
```
Experimental Evaluation

• Matrix Multiplication Example
  
  Matrix dimensions: \( nxn \), where \( n = 2, 3, \ldots, 7 \)
  
  Number of processes: \( numprocs = 1, \ldots, 10 \)

• Gauss-Jordan Elimination
  
  – Sequential
    
    Matrix sizes: \( mxn \), where \( m, n = 2, 3, 4 \)
  
  – Parallel
    
    Implementation and experimental runs in progress

Measured memory usage

  Main issue is the size of the State Vector;
  some runs use 85% of the entire machine memory

Distinguishing feature

  the common problem of state explosion of
  overshadowed by issues associated with the size of the State Vector
Experimental Results: Scalability of the Computation

Multiplication of $n \times n$ Matrices, log scale

Total memory usage (MB)

Number of Processes

$n=2$

$n=3$

$n=4$

$n=5$

$n=6$
Experimental Results: Optimization Options in SPIN

-DCOLLAPSE

![Graph of multiplication of nxn matrices on a log scale showing the impact of different numbers of processes on total memory usage](image-url)
Future Work

• Improvement of PROMELA models
  – Data structures
  – Incorporating C code
  – Theorem proving packages
• Using a different model checker, e.g. MOVer (MPI-Optimized Verifier)
• Exploring other non-trivial computational examples
Conclusions

• Deadlock
  – demonstrated applicability of abstractions
  – scaled somewhat reasonably – ability to do non-trivial sizes of matrices

• Computational correctness
  - sequential model capable of handling non-trivial sizes of matrices
  - using SPIN to create symbolic expressions and comparing these expressions for the parallel and sequential versions of the algorithm

• Initial experimental results are promising!
Gauss-Jordan Elimination: Procedure

Step 1. Locate the leftmost column (vertical line) that does not consist entirely of zeros

\[
\begin{bmatrix}
0 & 0 & -2 & 0 & 7 & 12 \\
2 & 4 & -10 & 6 & 12 & 28 \\
2 & 4 & -5 & 6 & -5 & -1
\end{bmatrix}
\]

Leftmost nonzero column
Gauss-Jordan Elimination: Procedure

Step 2. Interchange the top row with another row, if necessary, so that the entry at the top column found in Step 1 is different from zero

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\end{pmatrix}
\]

Interchange the first and second rows
Gauss-Jordan Elimination: Procedure

Step 3. If the entry that is now at the top of the column found in Step 1 is \( a \), multiply the first row by \( 1/a \) in order to introduce a leading 1.

\[
\begin{bmatrix}
1 & 2 & -50 & 6 & 62 & 28 \\
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\end{bmatrix}
\times \frac{1}{2}
\]
Gauss-Jordan Elimination: Procedure

Step 4. Add suitable multiples of the top row to the rows above and below so that all entries below leading 1 become zeros

\[
\begin{bmatrix}
1 & 2 & -5 & 3 & 6 & 14 \\
0 & 0 & -2 & 0 & 7 & 12 \\
0 & 0 & 55 & 0 & -15 & -29 \\
\end{bmatrix}
\]

2 times the first row was added to the third row
Background

• Distributed computing
  MPI, the Message-Passing Interface (C bindings)
• Symbolic execution of programs with non-trivial arithmetic
  – Matrix multiplication
    Multiplication of two square matrices;
  – Gauss-Jordan elimination
    Computing a reduced row-echelon form of any nxm matrix on n processes
• Model checking
  SPIN, Simple PROMELA INterpreter
Matrix Multiplication

Implementation in C using MPI:
• Written as an assignment for CS331 Programming Languages, University of Alaska Anchorage
• Structure and algorithm:
  – Files:
    matrices.txt – contains matrix size and values of all entries in initial matrices
    matrixmul.c – code;
  – Algorithm:
    1. Root process reads the dimensions (n) of the input matrices (A, B) and values of their respective entries
    2. Root process broadcasts n, A, and B to all processes
    3. Each process computes an entry of the resultant matrix represented by a new matrix C if its rank is equal to the remainder of the index of that entry (in row major order)
    4. The final result of the computation is accumulated in a matrix D on the root process by reducing the maximum of the corresponding entries in C from all the processes
Matrixmul.c

double *A, *B;       //two initial matrices
double *C;           //intermediate result
double *D;           //final result, D = AB

/* Read matrix dimensions from file*/
if(rank == 0){
    fscanf(inFile, "%d\n", &n);
    nElts = n*n;
}

/* Broadcast the dimension and the total number of elements in each matrix*/
MPI_Bcast(&nElts, 1, MPI_INT, 0, MPI_COMM_WORLD);
MPI_Bcast(&n, 1, MPI_INT, 0, MPI_COMM_WORLD);

/* Allocate memory for A, B, and C*/
A, B, C = (double *) malloc(nElts * sizeof(double));

/* Root process allocates memory for D and reads entries of A and B from file */
if (rank==0){
    D = (double *) malloc(nElts * sizeof(double));
    for(i=0; i<nElts; i++) {
        if (i%n==0) fscanf(inFile, "\n");
        fscanf(inFile, "%lf", &A, &B[i]);
    }
}

Matrixmul.c

/* Broadcast entries of A and B */
MPI_Bcast(A, nElts, MPI_DOUBLE, 0, MPI_COMM_WORLD);
MPI_Bcast(B, nElts, MPI_DOUBLE, 0, MPI_COMM_WORLD);

/* Computation */
for (i = 0; i<n; i++)
  for (j=0; j<n; j++) {
    C[(i*n)+j]=0;
    if (((i*n)+j)%size==rank) for (k = 0; k < n; k++) C[(i*n)+j] += (double)A[(i*n)+k] * (double)B[(k*n)+j];
  }

/* Reduce entries in C to the corresponding entries D on the root process*/
for (i = 0; i<nElts; i++)
  MPI_Reduce(&C[i], &D[i], 1, MPI_DOUBLE, MPI_MAX, 0, MPI_COMM_WORLD);
Properties

1. Freedom from deadlock

2. Correctness of computations
Modeling MPI Functions

`matrixmul.c` utilizes two MPI functions, both of which are collective:

- **MPI_Bcast** (void *buffer, int count, MPI_Datatype datatype, int root, MPI_Comm comm)
  
  *Broadcasts a message from the process with rank root to all processes of the group, itself included*

- **MPI_Reduce** (void *sendbuf, void *recvbuf, int count, MPI_Datatype datatype, MPI_Op op, int root, MPI_Comm comm)
  
  *Combines the elements provided in the input buffer of each process in the group, using the operation op, and returns the combined value in the output buffer of the process with rank root*

Based on the description in Siegel and Avrunin, *Modeling MPI Programs for Verification*, introduce another process for coordinating collective operations...
Coordinator Process

active proctype Coordinator() {
   /* Receive initial messages from all processes and confirm that all nodes agree on the type of communication in progress*/
   do
      :: chan_c[0]?im;
      i = 1;
      do
         :: i < nprocs -> chan_c[i]?cm; assert(im == cm); i++
         :: else -> i = 0; break
      od;
      if
         :: im == BCAST ->
            /* Interpret a broadcast message, receive a DONE message from every non-root node */
            do
               :: i < nprocs -> chan_c[i]?cm; assert(cm == DONE); i++
               :: else -> i = 0; break
            od
         :: im == REDUCE ->
            /* Interpret a reduce message, receive a DONE message from the root process */
            ... fi
      od
   od
}
Abstraction of C Code

Processes in PROMELA:

```c
active proc Root{...}
active proc [nprocs - 1] NonRoot{...}
active proc Coordinator{...}

/* Data channels from root to non-root nodes */
chan chan_from_root[nprocs]

/* Data Channels from non-root nodes to root */
chan chan_to_root[nprocs]

/* Collective communication channels from nodes to Coordinator */
chan chan_c[nprocs]
```

Diagram: 
- Coordinator
- NonRoot[0]
- ... 
- NonRoot[nprocs - 1]
- Root
Deadlock Model

• Message abstraction

\[ mtype = \{ \text{BCAST}, \text{REDUCE}, \text{DONE}, \text{DATA} \} \]

• Modeling possible need for synchronization of every send operation by implementing a non-deterministic choice

```plaintext
if ::1 -> empty(channel) ::1 fi
```
inline MPI_Bcast(datum, root){
    /* Initiate broadcast */
    chan_c[pid]!BCAST;
    if
        pid == root ->
        /* Send data to all procs */
        bcastData(datum, root)
    else ->
        /* Receive data from root */
        chan_from_root[pid]!datum
    fi;
    /* Notify coordinator upon completion*/
    chan_c[pid]!DONE;
    /* Account for a possibility of a blocking send */
    if
        !empty(chan_c[pid])
    then
        /* Notify coordinator upon completion */
        chan_c[pid]!DONE;
    fi
}

inline MPI_ReduceR(root){
    chan_c[0]!REDUCE; /* Initiate reduce */
    ... do
        i < nprocs ->
            if
                i != root ->
                /* Receive data from non-root procs */
                chan_to_root[i]!DATA
            else ->
                fi;
        ... od;
    chan_c[pid]!DONE; /* Notify coordinator upon completion */
}

inline MPI_ReduceNR(root){
    chan_c[pid]!REDUCE; /* Initiate reduce */
    chan_to_root[pid]!DATA; /* Send data to root */
    chan_c[pid]!DONE; /* Notify coordinator upon completion */
}
Deadlock Model

/* Root process */
active proctype Root() {
    MPI_Bcast(DATA, 0); /* Broadcast matrix dimension n */
    MPI_Bcast(DATA, 0); /* Broadcast n*n */
    MPI_Bcast(DATA, 0); /* Broadcast A */
    MPI_Bcast(DATA, 0); /* Broadcast B */
    /* Reduce the result */
    do :: i < n*n -> MPI_ReduceR(0); i++
       :: else -> i = 0; break
    od
}

/* Non-Root process */
active [nprocs-1] proctype NonRoot() {
    MPI_Bcast(DATA, 0); /* Receive n */
    MPI_Bcast(DATA, 0); /* Receive n*n */
    MPI_Bcast(DATA, 0); /* Receive A */
    MPI_Bcast(DATA, 0); /* Receive B */
    /* Reduce the result to the root node */
    do :: i < n*n -> MPI_ReduceNR(0); i++
       :: else -> i = 0; break
    od
}
Computing Model

- MPI Communication – the same as in the deadlock model
- Expressions
  - Representation
    Symbolic expressions are stored in byte arrays in prefix notation
    ```
    typedef Expn{
        byte length;
        byte array[maxExpnSize]
    }
    ```
  - Symbolic constants represent operators, e.g.
    ```
    #define 255 PLUS
    ```

Example:

\[ a_2 \cdot a_3 + a_4 \cdot a_5 \text{ in infix notation is equivalent to } + \cdot a_2 \cdot a_3 \cdot a_4 \cdot a_5 \text{ in prefix notation and to } 254 \ 2 \ 3 \ 254 \ 4 \ 5 \text{ in the model} \]
Computation Model

- Manipulation of Expressions
  
  inline addTo(a, b){…} results in a = a+b
  inline multBy(a, b){...} results in a = a*b

Example:

```
2 * 3 + 4 * 5 = 254 2 3 255 254 2 3 254 4 5
```
Computation Model

1. Symbolic computation is performed in parallel, generating a matrix of expressions on the root process

2. The root process does the symbolic computations sequentially

3. The root process loops through the two resultant structures checking that the two are the same via a set of assertions
Scalability: Computation

Multiplication of $n \times n$ Matrices, log scale

- $n=2$
- $n=3$
- $n=4$
- $n=5$
- $n=6$

Total memory usage (MB) vs Number of Processes
Optimization options in SPIN
-DCOLLPASE
Matrixmul summary

- **Model 1**: Freedom from deadlock
  Verified for 10x10 matrices with 4 processes or 4x4 matrices with 16 processes

- **Model 2**: Correctness of computation
  Verified that the correct result is produced for all possible executions for 3x3 matrices with 7 processes, ...
Gauss-Jordan Elimination

Common application finding a matrix inverse

\[
[A \ I] \equiv \begin{bmatrix}
    a_{11} & \cdots & a_{1n} & 1 & 0 & \cdots & 0 \\
    a_{21} & \cdots & a_{2n} & 0 & 1 & \cdots & 0 \\
    \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\
    a_{n1} & \cdots & a_{nn} & 0 & 0 & \cdots & 1
\end{bmatrix},
\]

where \( I \) is the identity matrix, to obtain a matrix of the form

\[
\begin{bmatrix}
    1 & 0 & \cdots & 0 & b_{11} & \cdots & b_{1n} \\
    0 & 1 & \cdots & 0 & b_{21} & \cdots & b_{2n} \\
    \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
    0 & 0 & \cdots & 1 & b_{n1} & \cdots & b_{nn}
\end{bmatrix} \cdot \begin{bmatrix}
    b_{11} & \cdots & b_{1n} \\
    b_{21} & \cdots & b_{2n} \\
    \vdots & \ddots & \vdots \\
    b_{n1} & \cdots & b_{nn}
\end{bmatrix}
\]

is then the inverse of \( A \) if such exists and \([A \ I]\) is in the reduced row-echelon form.
Gaussian-Jordan Elimination Algorithm

Repeat for every row of the matrix:

• Locate the leftmost column that does not consist entirely of zeros.
• Interchange the top row with another row, if necessary, so that the entry at the top of the column found in Step 1 is different from zero.
• If the entry that is now at the top of the column found in Step 1 is a, multiply the first row by $1/a$ in order to introduce a leading 1.
• Add suitable multiples of the top row to the rows above and below so that all entries above and below the leading 1 become zero.

Example:

\[
\begin{bmatrix}
0 & 0 & -2 & 0 & 7 & 12 \\
2 & 4 & -10 & 6 & 12 & 28 \\
2 & 4 & -5 & 6 & -5 & -1 \\
\end{bmatrix}
\rightarrow
\begin{bmatrix}
1 & 2 & 0 & 3 & 0 & 7 \\
0 & 0 & 1 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 1 \\
\end{bmatrix}
\]
Matrix Multiplication vs. Gauss-Jordan Elimination

Why is Gauss-Jordan Elimination harder?

• There is need to introduce branching, where conditions are functions of data
• The property is harder to express since there is no closed formula of the answer, again, consequence of data dependencies
Example of Symbolic Gauss-Jordan Elimination on a 2X2 matrix

A = \[
\begin{bmatrix}
a_{11} & a_{12} \\
a_{21} & a_{22}
\end{bmatrix}
\]

Case 1:
\[
a_{11} = 0
\]
\[
a_{21} = 0
\]
\[
a_{12} = 0
\]
\[
a_{22} = 0
\]
\[
\begin{bmatrix}
0 & 0 \\
0 & 0
\end{bmatrix}
\]

Case 2:
\[
a_{11} = 0
\]
\[
a_{21} = 0
\]
\[
a_{12} = 0
\]
\[
a_{22} \neq 0
\]
\[
\begin{bmatrix}
0 & 0 \\
0 & 1
\end{bmatrix}
\]

Case 3:
\[
a_{11} = 0
\]
\[
a_{21} = 0
\]
\[
a_{12} \neq 0
\]
\[
a_{22} \text{ free}
\]
\[
\begin{bmatrix}
0 & 1 \\
0 & 0
\end{bmatrix}
\]

Case 4:
\[
a_{11} = 0
\]
\[
a_{21} \neq 0
\]
\[
a_{12} \text{ free}
\]
\[
a_{22} \text{ free}
\]
\[
\begin{bmatrix}
1 & 0 \\
0 & 1
\end{bmatrix}
\]

Case 5:
\[
a_{11} \neq 0
\]
\[
a_{22} - a_{21} \cdot a_{12} / a_{11} \neq 0
\]
\[
\begin{bmatrix}
0 & a_{12} / a_{11} \\
0 & 0
\end{bmatrix}
\]
Approach to Verification

1. Construct both sequential and parallel models
2. Assume that the sequential model algorithm is correct
3. Show that for all input values given to the sequential algorithm, the parallel version will arrive at the same result.

Dealing with branching on data:

The sequential version executes the algorithm first, exploring both possible branches on values of matrix entries, that is

\[
\begin{align*}
a_{11} \neq 0 \quad \text{or} \\
a_{11} = 0
\end{align*}
\]

PROMELA:

\[
\begin{align*}
\text{if} \\
\rightarrow 1 \rightarrow a_{11} = 0; \ldots \\
\rightarrow 1 \rightarrow a_{11} \neq 0; \ldots \\
\text{fi}
\end{align*}
\]
Approach to Verification

Sequential model
1. accumulates a table of path conditions
2. passed on to the parallel code along with the generated expression for the answer

Parallel model
1. uses the path conditions of the table generated by sequential model and checks that the symbolic expressions are the same as the ones obtained sequentially
Gauss-Jordan Elimination Work in Progress

Implemented so far:

• Sequential version in C
• Parallel version in C using MPI
• Sequential version of the SPIN/PROMELA model that explores all possible executions and accumulates a table of expressions of path conditions

Remaining to implement:

• Parallel version of the SPIN/PROMELA model which takes the input data, table of path conditions, and sequentially computed resultant matrix
The End
\[ a = 2 \times 3 \]

\[ b = 4 \times 5 \]

\[ a + b = (2 \times 3) + (4 \times 5) = 254 \times 2 \times 3 \]
Matrix Multiplication

**Definition:**
The product $C(n \times m)$ of two matrices $A(n \times p)$ and $B(p \times m)$, is defined by

$$c_{ik} = a_{ij}b_{jk}$$

where $j$ is summed over for all possible values of $i$ and $k$. 

$$\begin{bmatrix}
  c_{11} & c_{12} & \cdots & c_{1p} \\
  c_{21} & c_{22} & \cdots & c_{2p} \\
  \vdots & \vdots & \ddots & \vdots \\
  c_{n1} & c_{n2} & \cdots & c_{np}
\end{bmatrix} = 
\begin{bmatrix}
  a_{11} & a_{12} & \cdots & a_{1m} \\
  a_{21} & a_{22} & \cdots & a_{2m} \\
  \vdots & \vdots & \ddots & \vdots \\
  a_{n1} & a_{n2} & \cdots & a_{nm}
\end{bmatrix} 
\begin{bmatrix}
  b_{11} & b_{12} & \cdots & b_{1p} \\
  b_{21} & b_{22} & \cdots & b_{2p} \\
  \vdots & \vdots & \ddots & \vdots \\
  b_{m1} & b_{m2} & \cdots & b_{mp}
\end{bmatrix}.$$
typedef Expression{ /* Symbolic expression in prefix notation */
    byte length;
    byte array[entrySize]
} 

typedef SArray{ /* Original matrices. integer symbols as entries*/
    byte length;
    byte array[matrixSizeSquared]
} 

typedef matrix{ /* Resultant matrix of expressions */
    Expression array[matrixSize2]
}

/* Root process */
active proctype Root() {
    ...
    /* D now has the result computed in parallel */
    /* Compute the result locally, store in matrix C */
    i = 0;
    j = 0;
    do
        :: i < n ->
            do
                :: j < n ->
                    k = 0;
                    do
                        :: k < n ->

mult(A.array[i*n + k], B.array[k*n + j], tmp);
addTo(C.array[i*n + j], tmp);
k++
:: else -> k = 0; break
od; j++
::else -> j = 0; break
od; i++
:: else -> i = 0; break
od;
/* Verify that D and C are the same */
i = 0;
do :: i < nElts ->
   assert(D.array[i].length == C.array[i].length);
   ii = 0;
do :: ii < entrySize -> assert(D.array[i].array[ii] == C.array[i].array[ii]);
      ii++
:: else ii = 0; break
od;
i++
:: else -> i = 0; break
od;
}
Typical Model Checking Architecture

- Program (C, MPI)
- Program Model (PROMELA)
- Model Checker (SPIN)
- Property

Abstraction:
- property verified
- counter example (execution trace) produced