Challenges in Constraint Programming for Hardware Verification

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Who are we?

- Verification Technologies Department (Simulation Based verification)
  - Part of IBM Research
  - Center of competence for verification technologies in IBM
  - Over two decades of experience in development of verification technologies for simulation based verification
    - Core level, System-level, Unit level
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Wesam Ibraheem
The significance of functional verification

✦ Roughly 70% of the design effort (time, resources, …) is invested in functional verification
✦ Industry practice: verification == over 90% simulation based verification

✦ A design re-spin may cost many millions of $
 ✦ Masks
 ✦ Person-month
 ✦ Time-to-market
✦ Typically 3-4 re-spins for complex designs (processors)
Key Technologies for Processor Verification

- **Genesys-Pro**
  - State-of-the-art test generator for full processor and multi-processor verification
  - Used by all IBM processors and licensed to external companies
    - Adaptable to any architecture
    - Applied in Power, zArch, ARM, and others
- **FPGen**
  - Dedicated generator focused on floating point verification
- **XGen**
  - A test generator for verification of systems
- **ThreadMill**
  - Post-Silicon and emulation exerciser
Content

- Part 1: GPro - Test generator for microprocessors
- Part 2: CSP characteristics and challenges
- Part 3: PRB – The new CSP approach
Part 1

GPro:
Test generator for microprocessors
Genesys-Pro: Model-Based Test Generation

- **Generic** architecture-independent test generation engine
- External formal and declarative **architecture description**
- **Behavioral simulator** used to predict instruction execution results
- Graphical User Interface to define **generation directives**
Resulting test case

INITIALIZATIONS: DATA MEMORY
D 06FFFFFA4  4ADDEB16

INITIALIZATIONS: REGISTERS
R R5             050C4340
R R6             09801460

INSTRUCTIONS
I 230DB4B4 46610FE7 * STRAL R6, [R15, -R6, ASR #0x02] * VA=FEADFA4 (New) RA=06FFFFFA4
I 230DB4B8 B3E77FEA * BAL 0x030D538C //0x7FE7B3 * VA=030D538C (New) RA=4ECBC38C
I 4ECBC38C 0E50BDE7 * LDRAL R5, [R13, +R14]! * VA=40BC6FA4 (Used) RA=06FFFFFA4

RESULTS: REGISTERS
R R5             09801460
R R6             09801460

RESULTS: DATA MEMORY
D 06FFFFFA4  60148009
Generation scheme – user view

1. Choose the **next instruction** to generate, according to:
   - Test template definition (test’s specification)
2. **Generate** instruction
   - Initialize resources as required
3. Call **reference model** to simulate instruction
4. Repeat until all test template statements generated
Random stimuli generator

System model:
- What’s valid
- What’s interesting

User requirements

Generate N tests

Constraint Satisfaction Problem

Random stimuli generator

CSP Solver

N distinct tests
- Valid, interesting
- Satisfy user requirements
GenesysPro system input: test template basics

- Instruction as the basic building block
- Full control over instruction properties:
  - Data, Address, Length,…
- A hierarchy of higher level statements
  - Select: weighted random choice
  - Repeat
  - Sequence

Test template

Sequence

Select

Repeat x10

Weight: 70
Load word
Target reg: {R3,R5}

Weight: 30
Store half-word
Addr: 0x12??
Source data: [-2, 2]

Add
Test template: testing knowledge and directives

- **Directives** as 'volume knobs' to control TK characteristics
- Testing knowledge also affects the test by default
- Directives present in the test template take precedence
- Scope based influence

```
Test template
Sequence
Repeat x10
Select
Load word
Store half-word
Add

Default TK

Address Collision: 65%

Cache hit:20
Cache miss:80
```
Instruction model

Object-oriented ontology language with a focus on constraint modeling
The concept of generic testing knowledge

- A set of mechanisms that aim at **improving test-case quality**
- Capitalize on **recurring concepts**
- The basic mechanism: non-uniform random choice
  - Bias towards ‘interesting’ areas
- Affects all generated test-cases
  - But can be controlled by users
- Examples:
  - Resource collisions
  - Translation table entry reuse
Testing knowledge example - placement

- A storage partition is a contiguous piece of memory
  - L2 cache line, page, word, half-word...
- Four types of events

- Boundary
- Alignment
- Crossing
- Vicinity
Why CP?

- CP enables requests coming from **different resources**
- CP gives the option to constraint **results**
- CP solvers enable approximation of **uniform coverage**

- The microprocessor **specification** is written declaratively
  - Easy translation into constraints
  - non-linear constraints

- Mandatory and **bias** (not mandatory) requests
Why CP?

Constraints originate from three sources

1. **Validity of the stimuli**: Constraints defined by the specification
2. **Verification task**: Constraints defined by the user
3. **Bias towards interesting tests**: Soft constraints defined by domain experts

**Validity**: Complex EA to RA translation

**Effective Address**: 0xB274FAB_0DBC0000

**Real Address**: 0x0002FFC5_90A4D000

**User**: EA aligned to 64K RA in some corner memory space

**Expert knowledge**: Reuse cache row
Not just IBM

- Constraint satisfaction is the basis for modern stimuli generation across the industry
- 42nd DAC:
  - The largest conference of the EDA industry, 6000 participants
  - A full-day tutorial about constraint satisfaction for stimuli generation
- A typical industrial advertisement:
  
  "Constraint-Driven Test Generation
  With Specman Elite's constraint-driven test generation, you can now automatically generate tests for functional verification. By specifying constraints, you can quickly and easily target the generator to create any test in your functional test plan …"
Part 2

CSP characteristics and challenges

See also:
E. Bin, R. Emek, G. Shurek, and A. Ziv,
Using constraint satisfaction formulations and solution techniques for random test program generation,
IBM Systems Journal 41, 2002
Random Solution

Requirement:

- Find many random, uniformly distributed, solutions of the same CSP
  - Many different tests from the same template
  - As opposed to one, all, or 'best' solution
  - Motivation: Test different computation paths of the microprocessor

Solution:

- Uniform solution distribution is approximated by random variable and value ordering

See also: Dechter et al., AAAI 2002
Huge domains

Requirement:
- The domain of many variables is $2^{128}$
  - Example: address space
  - In conjunction with arithmetic, bit-wise, and other types of constraints
  - Representation and operations on sets becomes an issue

Solution:
- Inaccurate representation (over approximation)
- Using also bit-vectors representation
Domain (set) representation example: bit-vectors

- All the addresses such that:
  - \( \text{addr} = \text{base} + \text{displacement} \) : architectural
  - \( \text{addr}[3:6] = 01\times1 \) : cache line
  - \( \text{addr} \in [0x20000000 : 0x10FFFFF] \) : memory space

- 'Masks' (bits vector) representation:
  - \( 0b01\times1 \rightarrow 0b0101, 0b0111 \)

- Exponential explosion
  - \( 01010101 + 0x0x0x0x0x \)
    \( \{10101010, 01101010, 10011010, 01011010, 10100110, 01100110, ..., 10010101\} \)
Hierarchy of constraints

Requirement:

- Different priority of constraints
  - Mandatory: test case validity
  - Non-mandatory: makes the test 'interesting'
  - Multiple levels of soft constraints – according to level of interest

Solution:

- Modeler specifies the constraint priority
- In each MAC, Mandatory constraints propagate first. Then one bias, mandatory constraints again, …
Coupled CSPs

A challenge:

- Cannot generate all instructions simultaneously
  - Instructions’ semantics is not modeled
  - Problem is too large
  - Constraint propagation computationally hard

A Partial Solution:

- Instructions are generated one at a time, and then executed by an ISS (Instruction Set Simulator)
- But … Instruction 3 may require a specific configuration
Conditional CSP

A challenge:

- Parts of the problem’s variables and constraints should not exist in the solution

A Solution:

- A tree representation. A node may have an ‘exists’ Boolean variable.
- Constraints within the existence node work as long as the ‘exist’ variable is not false.
- External variables have a shadow. The shadow var is synchronized with the real one when the exists variable becomes true.

See also: F. Geller and M. Veksler,
"Assumption-based pruning in conditional CSP", CP 2005
External variables (Remote)

A challenge:

- Some CSP variables cannot be represented as a set of discrete values
  - In the solution, the variable is not a single element
  - The variable is shared in several CSPs
  - Example: content of memory

A Solution:

- The engine holds a variable having no domain.
- The relations communicate with the database during propagation
- Relations mark the propagated variables as 'modified', so the engine knows which other propagators to call.
Run time performance

**Requirement:**

- Generation of a test should not take more time than its simulation time

**A Solution:**

- Instructions are generated one at a time
- Similar problems are cached and reuse
Our major CSP solver

- **GEC**
  - Systematic, based on MAC-3
  - Since 1995, many person-years invested
  - Finite domain set libraries: “PD” (**primitive domains**)
    - Bool, int, bit-vector, object, string
  - Generic **expression propagator (ERP)**
    - Given a first order logic expression over variables, creates a propagator
  - Interfaces for user-defined **C++ propagators**
  - Arc-consistency on **conditional problems**
  - Support application specific CSP variables (**remote** variables)
  - Written in C++
  - designed to be generic
    - i.e., not specific for verification

See also: IAAI 2006, AI-Magazine 2007
New challenges coming from the hardware

- More complex **micro-architecture**
  - Example: **SMT** (Simultaneous Multi Threaded)
  - More directed scenarios required
  - More requirement on inter instruction constraints

- More complex **architectures**
  - Example: Translation
  - Complex CSP, solving issues

- **Virtualization**
  - Translation CSP problem replicated
  - A scalability issue
Status for 2010

- ERP: the declarative constraints language
  - No access to generator’s internal values
  - Just primitive operators
  - Insufficient expressiveness

- Many constraints are written in C++:
  - C++ code produces a better run-time performance

Maintenance cost and modeling new designs become an issue
Why propagators C++ coding is not recommended

- Less **readable**
- Much more **lines** of code
- No **reuse**
- Hard to **maintain**
- Hard to **debug** (log file does not show the semantics)
- Does not enable **composition** of operators
- Some times written with just **partial propagation**
- Much more **time** to code it
- The CSP engine sees it as a **black box**
Part 3

PRB:
The new CSP approach
In short

- PRB address the shortcoming ERP of the
  - Allow greater **expressive**
  - Allow seamless **integration** with the application
  - Support **random** decisions in non-mandatory constraints
  - Built in solving **heuristics**
PRB. PRopagator Builder. Principles:

- **Primitive types:**
  - New primitive types
- **Constraints**
  - Constraints are written *declaratively* (not in C++)
  - Many new operators
  - Operators can be composed
- **Macros**
- **Interface:**
  - PRB communicates with the application
  - Application can configure PRB
- **Solving:**
  - Generic management of representation explosion problem
  - No modeling of propagation ordering
  - Semantics based variable and value ordering
Example: Direct access

```c
GP_STATUS GP_MATCH_LPIDR_VALUE(PD_BitStream &RS)
{
    TRY (GP_MATCH_LPIDR_VALUE)
    PD_BitStream word0, LPIDR;
    RS.GetSubField(0, 31, word0);
    static ROI_ObjId LPIDR_ID = REL_Kernel::GetMnemonicResourceId("LPIDR");
    REL_Kernel::GetRegisterContents(LPIDR_ID, LPIDR);
    Intersect(word0, LPIDR, "GP_MATCH_LPIDR_VALUE");
    Intersect(LPIDR, word0, "GP_MATCH_LPIDR_VALUE");
    RS.SetSubField(0, word0);
    if (RS.IsEmpty()) return GP_EMPTY;
    return GP_EXACT;
    CATCH
}
```

---

```
PRB_MATCH_LPIDR_VALUE:
    subField(data,32,63) = resources.LPIDR
```
constraint ERP: ERP_MaskAligned
  (Aligned_Addr: bitstream,
   Unaligned_Addr: bitstream,
   in Alignment: integer
   %FormalFacet:: < precondition: SingletonPrecondition, range: < > >
  )

["let AlignmentMask (PD_Int): PD_BitStream {"n
  (2) : 0xFFFF_FFFF_FFFF_FFFE,
  (4) : 0xFFFF_FFFF_FFFF_FFFC,
  (8) : 0xFFFF_FFFF_FFFF_FFF8,
  (16) : 0xFFFF_FFFF_FFFF_FFF0,
  (32) : 0xFFFF_FFFF_FFFF_FFE0,
  (64) : 0xFFFF_FFFF_FFFF_FFC0,
  (128): 0xFFFF_FFFF_FFFF_FF80,
  (256): 0xFFFF_FFFF_FFFF_FF00
  } ! A: PD_BitStream « 0XXXXXXXXXXXXXXXXXX »,
  B: PD_BitStream « 0XXXXXXXXXXXXXXXXXX »,
  C: PD_Int « {2, 4, 8, 16, 32, 64, 128, 256} »
];

{a in A;
 b in B;
 c in C;
	a = b bit_and AlignmentMask(c);
}"n"n"n"n
instance PRB_MaskAligned: PRB_GeneratorMacro = <
  parameters: "Aligned_Addr, Unaligned_Addr, Alignment",
  body: "Aligned_Addr = maskLsbField(Unaligned_Addr, log2(Alignment)-1, ZERO)";"n"
Constraints types

- All the following types reduce values from variables.

- **Propagator**
  - A deterministic logical / arithmetical algorithm.
  - Reduce values that do not have a support.
  - Used within MAC algorithm

- **Restrictor**
  - A non-deterministic logical / arithmetical algorithm.
  - It draw values
  - Used within MAC algorithm

- In addition to the priority of the constraint (mandatory / bias)
Operators wealth

- The more operators in the language
  - The modeling is shorter
  - More readable
  - Better propagation
  - Better run-time
Better propagation for higher level operators

Option 1: \( b = \text{concat}(\text{subField}(a, y+1, 63), \text{subField}(a, 0, x-1)) \)

Option 2: \( b = \text{pullOutSubField}(a, x, y) \)

Option 1 produces weaker propagation:
1. A delay: When \( x, y \) are not a single element
2. Tightness: ‘concat’ collect too many elements

\[
\begin{align*}
a &= \{ 0011, 1100, 0111, 0100, 0000 \} \\
\text{option 1: } b &= \{0011, 0000, 0111, 0100\} \\
\text{option 2: } b &= \{0011, 0100\}
\end{align*}
\]
Old primitive set types

- Integers
- Boolean
- String
- Enums
- Bits vector
New primitive types

- Bits vector now have different formats
  - Plain bits
  - Unsigned integers
  - Signed integers *
  - Decimal representation
  - Floating point representation *

- Interval
  - Each interval holds two primitive sets for ‘start’ and ‘length’

- Dates

* Not done yet
Operators

- To have a feel of the operator library, we will see different operators
  - Just examples (there are more)
  - We will not understand the semantics of all of them (a quick session)
  - The syntax is not the issue
Intervals geometric operators: examples

- x before y
- x conscutivesTo y
- x adjacent y
- x crossesBeyond y
- x crosses y
- x sameBoundary y
- x overlaps y
- x contains y
- x shorterThan y
Global constraints: examples

- allDiff
- sumOf
- numOf
- exist
- collect
- select
- forAll
- forEach
- minOf
- maxOf
Properties of global constraints

- **Similar syntax** for all global constraints
- Formats:
  - Using vectors:  `forAll(i, 0, 7, vec[i].size > 0)`
  - Using objects:  `forAll(i, homes({employes}), i.salary > 20000)`
  - Using items:  `forAll(i, items({from, to}), shape.i < 100)`

- Conditions: Optional
  - `allDiff(i, homes({roads}), i.city != NY, i.name)`

Italic represents a PRB reserved word
Fields operators: examples

- carry
- concat
- subField
- extend
- maskField
- setField
- overflow
- pullOutSubField
- sameLsb
- numLsbBits
- Bitwise operations
Square Parentheses [ ] operators

- Direct access to a field of a known register
  - resources.MSR[TR]
    - is equivalent to
      - subField(resources.MSR, 4, 6)
  - The application informs PRB about all the known register fields

- The indirect operator
  - vec[x+3] = y
    - both x and y are CSP variables
The triple operator

- $x = (\text{condExp} \ ? \ \text{thenExp} : \text{elseExp})$

  This operator was found essential.

- $x = (\text{cond1} \ ? \ \text{then1}, \text{cond2} \ ? \ \text{then2}, \text{cond3} \ ? \ \text{then3}, .... \text{condN} \ ? \ \text{thenN} : \text{else})$
Boolean Operators

- `memberOf`
- `table`
- `positive`
- `negative`
- `zero`
- `find`

```c
b = table(x, y, z,
    {
        ( << 0b0 >>, UBool, UBool) : false,
        ( << 0b1 >>, UBool, false) : false,
        // ( << 0b1 >>, false, true ) : illegal
        ( << 0b1 >>, true, true) : true
    });
```

The tuples of the table can be generated in run time.
Restrictors Operators

- choose
- maxValue
- minValue
- randomBool
- randomMSBValues
- randomWeightedNumber
- randomWeightedValue
- randomNumber
- randomValue

These operators are legal just in non-deterministic constraints
Homes: background

- An application’s class. Inherits from PRB_Home.

- Includes (optionally):
  - Variables (inherits from PRB_Variable)
  - Constraints
  - Sub homes

- The application can add any data members / methods

- The home serves PRB during constraint hatching:
  - FindVar()
  - GetImmediateValue()
  - GetHomesGivenType()
Interface: PRB <-> Application

Propagators creation

- The application creates a tree of ‘home’s
  - Each home holds CSP variables, Constraints and sub homes

- Propagators creation
  - This interface enables sending expressions with unknown number of variables

Application (home) — Hatch — PRB

- FindVar, GetImmediate, ...
- Returns propagator(s)
Interface: PRB <-> Application Configuration

- Reserved words
- Max number of masks per variable
- Register fields
- Table’s tuples
- Macros
- Depth of conflict detection
- … and many more
Over approximation

- PRB over approximates the variable’s content
- Requirements:
  - **Reduce** the number of masks to the requested level
  - Do not insert values that were not in the variable’s domain when entering the propagator
  - Insert as few values as can

- Partial solution
  - While the number of masks is too many
    - Find two similar masks (heuristics)
    - Combine the masks
    - Reduce other masks that contained in the new one

\[ 0\text{bxxxxx1} \quad 0\text{bxxx1x} \quad 0\text{bxxxxx} \]
Conflict Detection

- $a > b$
- $a < b$

Constraints contradiction should be handle specifically since regular MAC with large domains does not cope with it efficiently.

Our solution: instrumentation
- Insert an auxiliary variable $v$
- convert the constraints
Conflict Detection: examples

**Original:**
(a > b) and (b > a)

**Instrumented:**
(v_{a,b} > 0) and (v_{a,b} < 0) and (v_{a,b} > 0 \leftrightarrow a > b) and (v_{a,b} < 0 \leftrightarrow a < b)

**Original:**
(a > b) and (x=1) \rightarrow (a < b))

**Instrumented:**
(v_{a,b} > 0) and (x=1) \rightarrow (v_{a,b} < 0)) and (v_{a,b} > 0 \leftrightarrow a > b) and (v_{a,b} < 0 \leftrightarrow a < b)
Semantics Variable and Value ordering (heuristics)

- When the semantics of the constraint is not a black box, it can be used for variable and value ordering.

- Two methods:
  - Static – partial ordering is defined before solving starts
  - Dynamic – ordering is defined during solving time

- Both methods neither use the number of values in a domain nor the constraints graph.
Static Semantics Variable Ordering

1. Variable V is selected randomly as a candidate to be instantiated
2. If all the variables Vs that V depends on have a single value, return V otherwise, choose randomly a variable from Vs and go to 2.

Comments:
1. If variables’ cycle is exposed, the variables in the cycle do not returned.
2. Work on fields granularity.

Characteristics:
- Automatic
- Sensitive to the way the user writes the constraints
- Works in causal CSP networks
Static Semantics Variable Ordering: examples

- Equal operator at the constraint’s tree root:
  \[ a = b + c \]
  \[ a \text{ depends on } b, c \]

- ‘imply’ operator at the constraint’s tree root:
  \( (a > 7) \rightarrow (b > c) \)
  \[ b, c \text{ depends on } a \]

- Fields granularity
  \[ \text{subField}(a, 2, 3) = \ldots \]
  \[ \text{just the two bits of } a \text{ are depended} \]
Dynamic Semantics Variable and Value Ordering
Motivation

<p>| | | | | | | | |</p>
<table>
<thead>
<tr>
<th></th>
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<td>92</td>
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</tr>
</tbody>
</table>

The domain of \( v[i] \) is \([0, 100]\)

- \( \text{numOf}(i, 0, 9, v[i] = 0) > 0 \)
- \( \text{numOf}(i, 0, 9, v[i] = 1) > 0 \)
- \( \text{numOf}(i, 0, 9, v[i] = 2) > 0 \)
- \( \text{numOf}(i, 0, 9, v[i] = 3) > 0 \)  \(\checkmark\)
- \( \text{numOf}(i, 0, 9, v[i] = 4) > 0 \)
- \( \text{numOf}(i, 0, 9, v[i] = 5) > 0 \)
Dynamic Semantics Variable and Value Ordering

- During regular propagation, when a propagator has multiple ways to be satisfied, it registers itself.
- During variable ordering:
  - Choose one of the registered propagators
    - Last one
    - Random one
  - Invoke the propagator in ‘ordering’ mode
  - When the propagator has multiple ways, it chooses one of them and satisfies it.
    - Last way
    - Random one
  - A variable does not change the real domain, but works on a copy
  - The variables that were copied (and their new domain) are the suggestion.
Wrap up

- Simulation is still the main platform for hardware verification
- Biased random test generation is widely used in the industry
- CSP is the major technique used for generating tests
- Architectures and micro-architectures enforce new CSP techniques
  - Modeling languages
  - Domain representation
  - Variable and value ordering
  - CSP debug methods
Thank you