

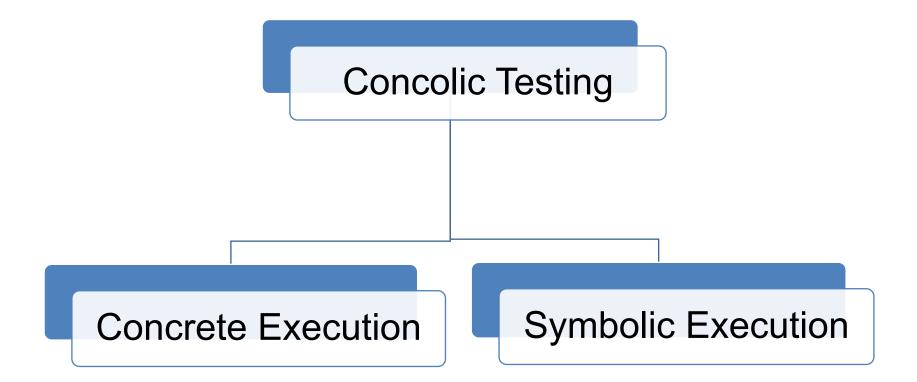
## Efficient Concolic Testing of MPI Applications

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# It Is Popular



- > Programming languages:
  - > Binary machine code, C, Java, and JavaScript.
- > Application types:
  - > web applications, sensor network applications, Unix utilities, database applications, and embedded software, GPU programs, image processing software, and so on
- > Various tools:
  - > KLEE, DART, SAGE, PEX, jCute, CREST, Jalangi, etc.

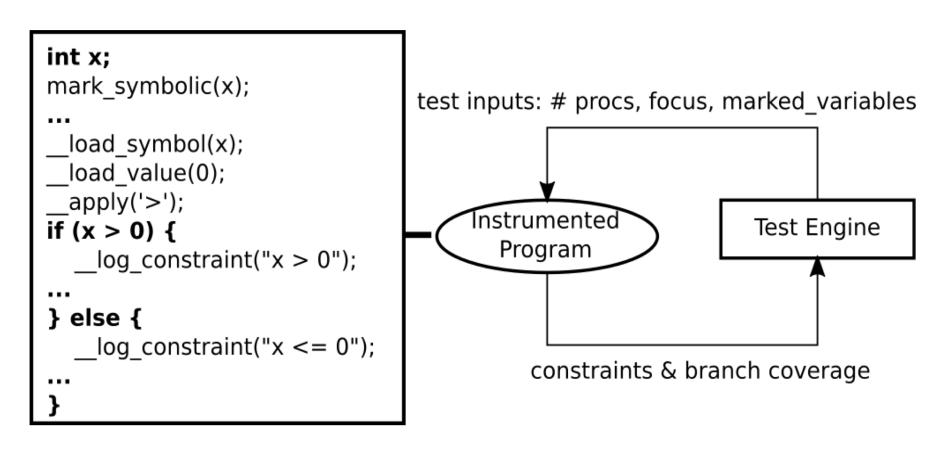
#### COMPI [IPDPS 2018]



- > COMPI is a concolic testing tool for MPI programs with following major features:
  - > Deals with basic MPI semantics
  - > Deals with high testing cost caused by input values, parallelism, and loops
- COMPI achieves 69-86% branch coverage within a few hours

## **Concolic Testing**

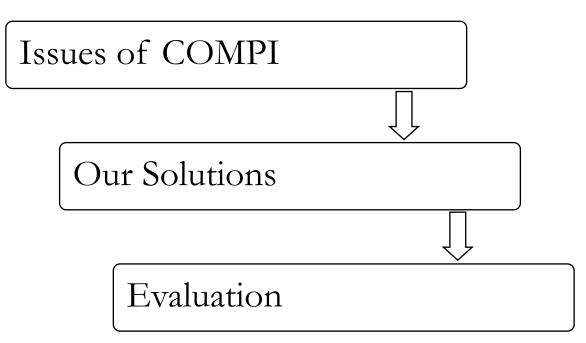




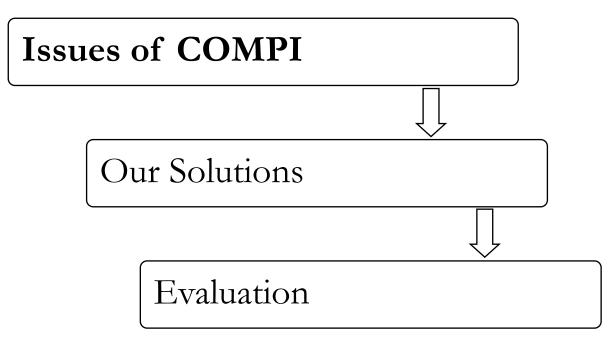
#### (1) Instrumentation

(2) Iterative testing









#### **Issues of COMPI**

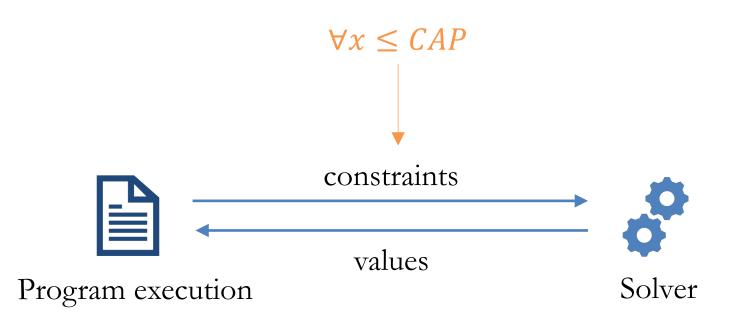


- > Input generation does not guarantee cost-effective testing
- > Floating point data types and operations are not supported

#### Issue I

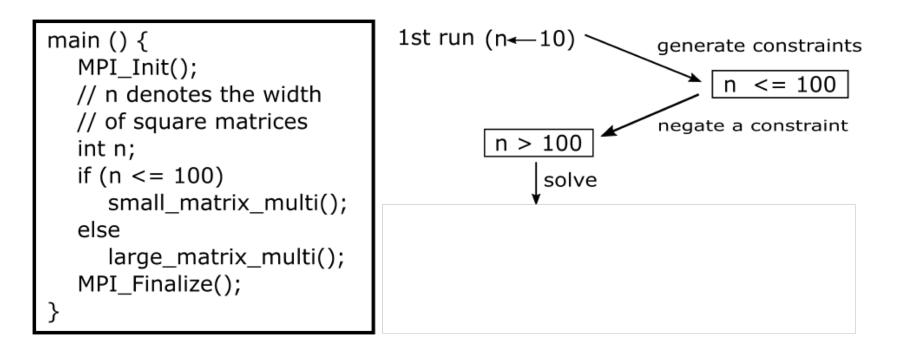


- > Larger input values  $\rightarrow$  Longer execution
- Solution of COMPI: Input Capping





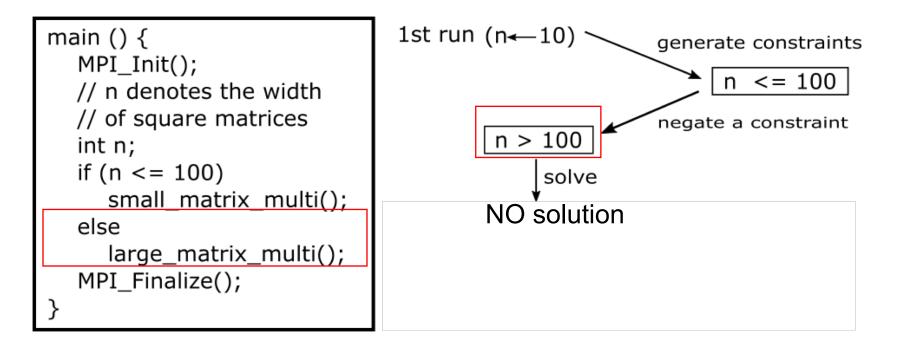




An MPI program performing matrix multiplication.



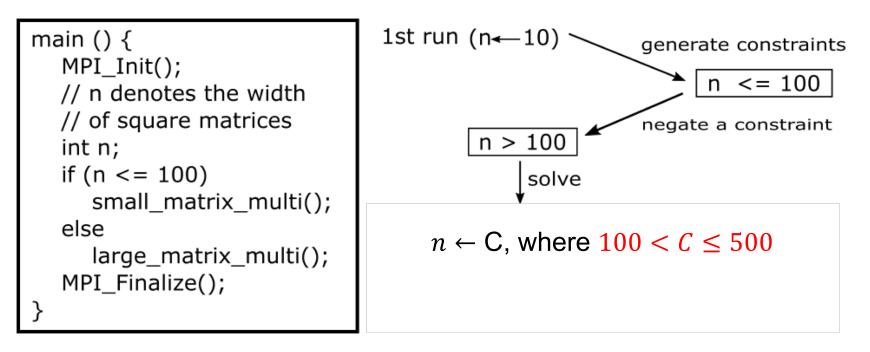




 $CAP = 50 \rightarrow$  Fail to cover the *else* branch

**Big Cap** 

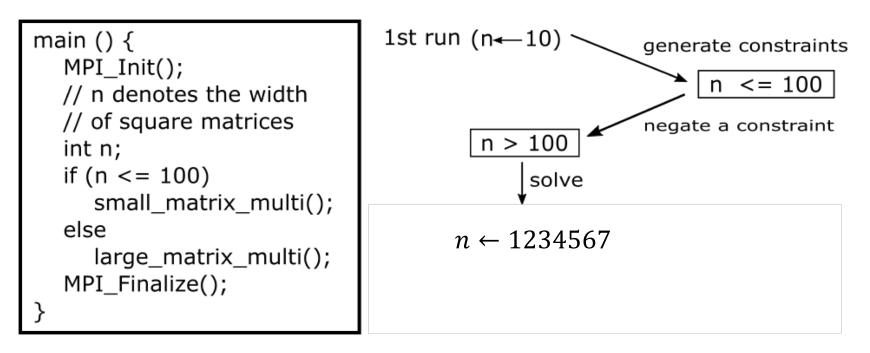




 $CAP = 500 \rightarrow$  High testing cost

## No Input Capping





No Capping  $\rightarrow$  Execution failure

#### Issue II



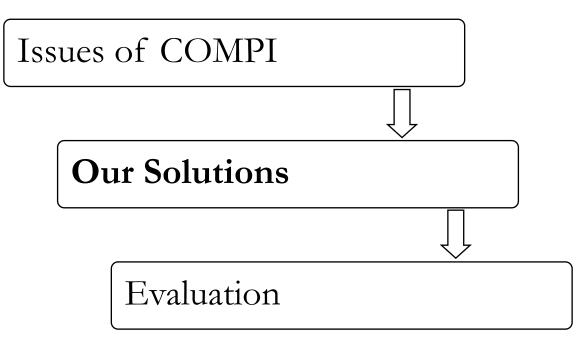
 Floating-point data types and operations are not supported



```
main () {
    ...
    int a; float b;
    COMPI_int(a);
    ...
    if (b > 1.1) f1();
    else f2();
    float c = a * 1.1;
    if (c > 2) f3();
    else f4();
}
```

- > Unable to mark *b* 
  - Fixing it to a value  $\rightarrow$ either *f1* or *f2* could not be explored
- > Unable to record a \* 1.1
  - > Symbolic representation of c is not existing  $\rightarrow$  either *f*<sup>3</sup> or *f*<sup>4</sup> could not be explored





#### **Our Solutions**



- > Input tuning  $\rightarrow$  cost effective testing
- > Floating-point extension → exploration of branches related to the use of floating-point arithmetic

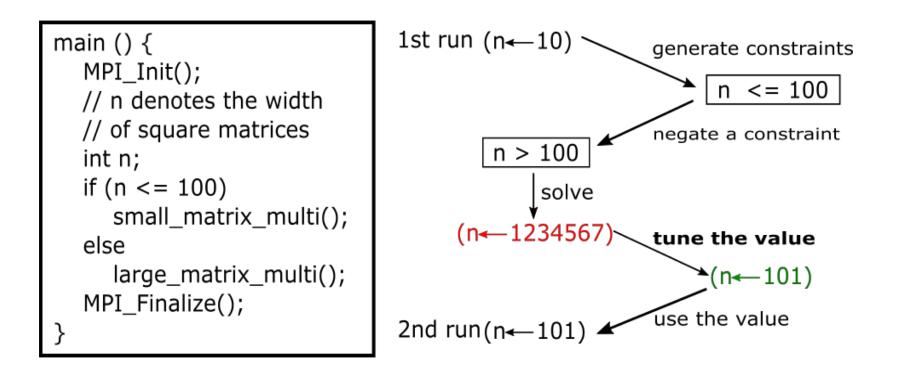
#### **Our Solutions**



- > Input tuning  $\rightarrow$  cost effective testing
- ➤ Floating-point extension → exploration of branches related to the use of floating-point calculations

# Input Tuning





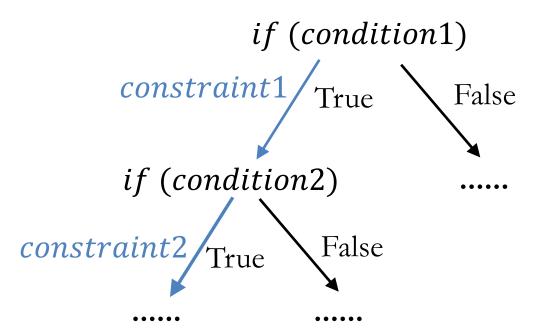
#### **Input Tuning**



Binary search of *upper* in (0, 1234567) satisfying:  $\{n > 100\} \cup \{n \leq upper\}$  is solvable AND  $\{n > 100\} \cup \{n \leq upper - 1\}$  is unsolvable Tuning  $n \leftarrow 1234567$  $n \leftarrow 101$ Tuned solution Solution by solver for  $\{n > 100\}$ 



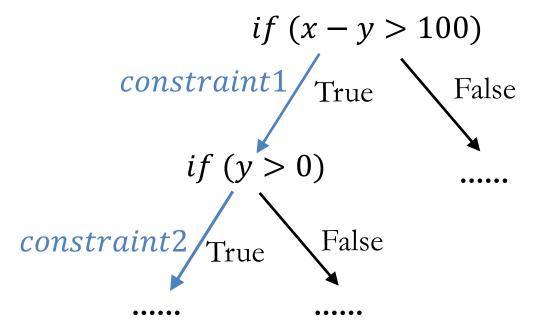
## Input Tuning for Multi-variable Multi-constraint Case



Need to solve {*constraint1*, *constraint2*}



## Input Tuning for Multi-variable Multi-constraint Case



Need to solve  $\{x - y > 100, y > 0\}$ 

# Input Tuning for Multi-variable **Multi-constraint Case** $\{x = 4321, y = 1234\}$ Stage I Tuning ${x = 4321, y = 1234, x \le upper1, y \le upper1}$ $min\{upper\} = 102$ Stage II Tuning $\{x = 4321, y = 1234, x \le upper1, y \le upper2\}$ $min\{upper2\} = 1$ ★ {x = 102, y = 1} Tuning for $\{x - y > 100, y > 0\}$

## Input Tuning -- Summary



- > Stage I avoids too large values being generated for ALL variables appearing in dependent constraints
- Stage II ensures the smallest value is generated for the SINGLE variable appearing in the target constraint based on Stage I

#### **Our Solutions**



- > Input tuning  $\rightarrow$  cost effective testing
- > Floating-point extension → exploration of branches related to the use of floating-point arithmetic

## **Floating-Point Extension**



- > Two floating-point data types supported: *float*, *double*
- > The extension adopts the design methodology of symbolic reasoning for integers
  - > Instrument floating-point operations
  - > Records only *linear constraints* 
    - Non-linear constraints are simplified using concrete values, e.g., x \* y is recorded as C \* x with C being the concrete value of y

#### **Floating-Point Extension**



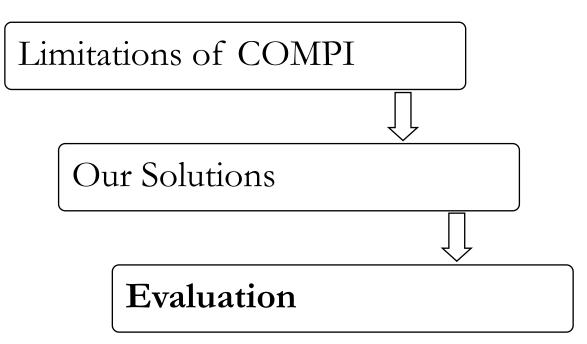
- > Two solvers: Real v.s. Float
  - > Accuracy: Real < Float
  - Solving speed: Real > Float

Real v.s. Float based on 100 iterative tests of a synthetic program that compares expression *e* with constant 0.

<i>e</i> =	x	x + y	x + y + z
Cost (float)	31.4s	75.0s	91.2s
Cost (real)	8.2s	8.1s	8.2s

3.8-11.1X faster

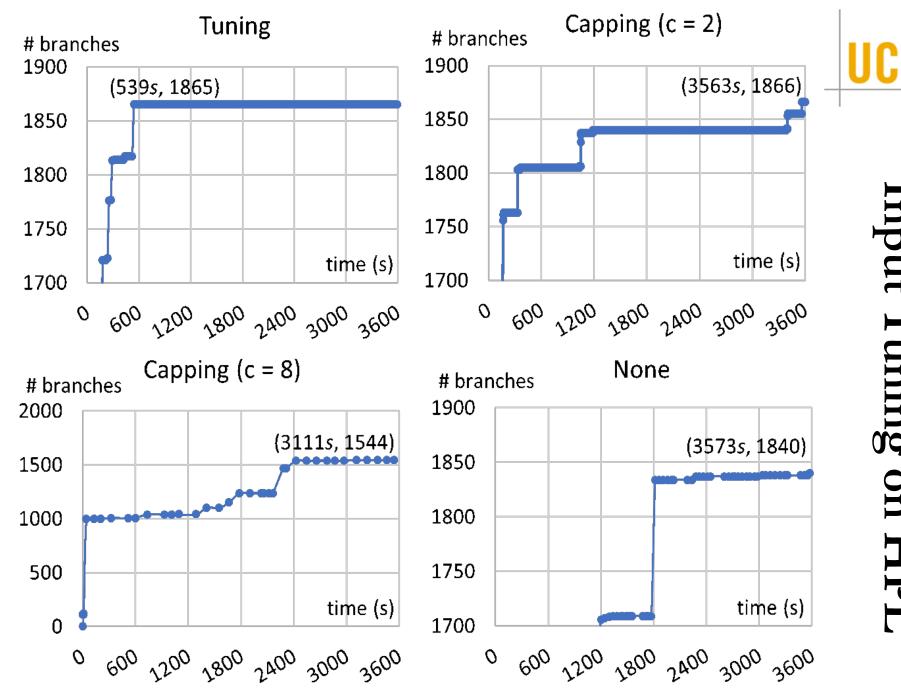




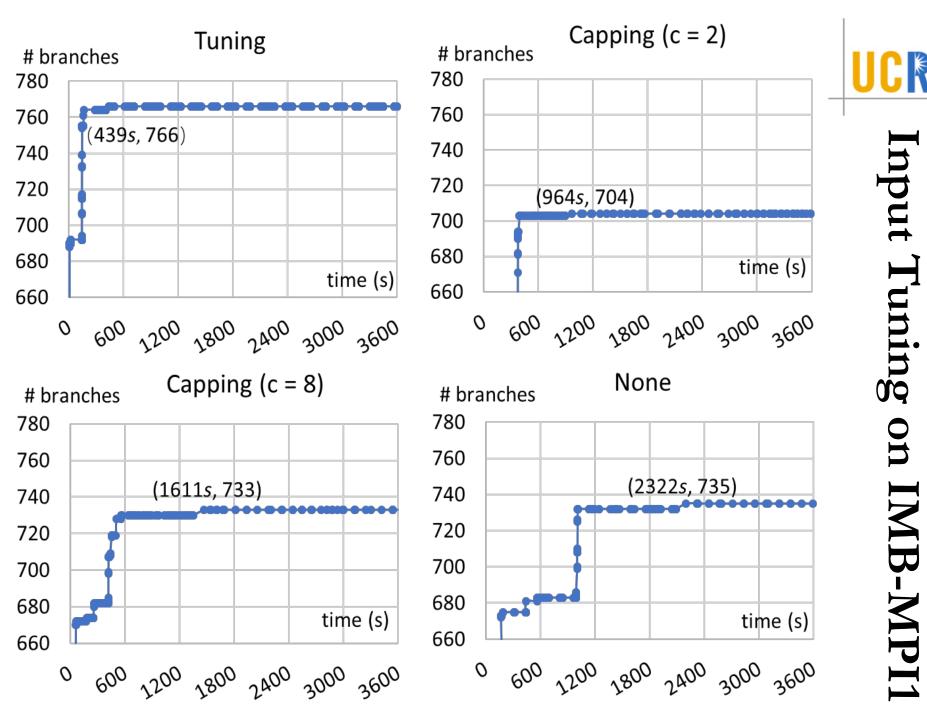
#### Evaluation

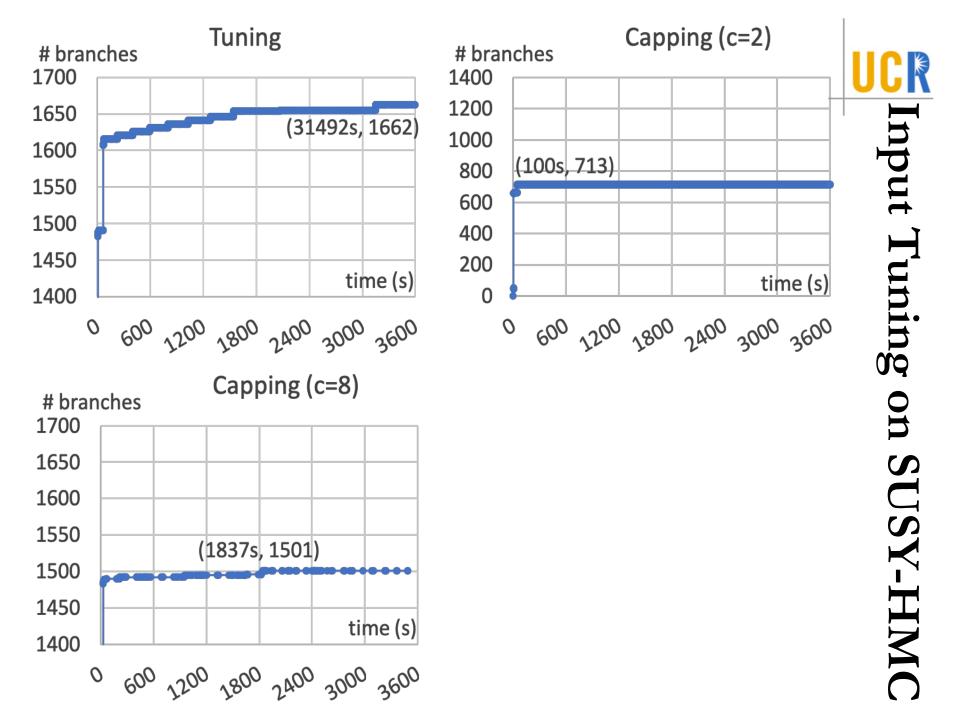


- Input tuning is evaluated using HPL, IMB-MPI1, and SUSY-HMC
- > Floating-point extension is evaluated using SUSY-HMC
- > One hour testing at each configuration
- > Initial input values are 1 for all variable in the first test
- > In the evaluation of input capping, we selects the same cap for all variables



Input Tuning ( on HPL





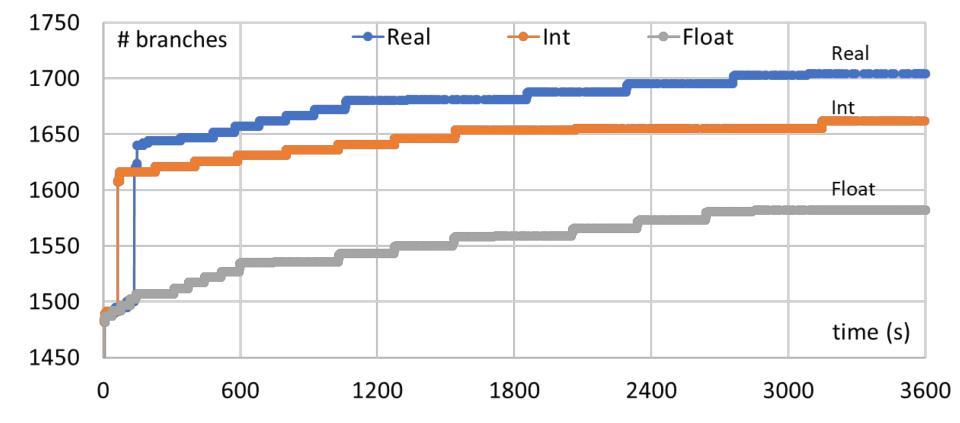
# Input Tuning



- ▶ 10-minute coverage (input tuning) ≥ 1-hour coverage (other methods)
- SUSY-HMC: 1-hour coverage (input tuning) is about
   1.2-2.3X higher than 1-hour coverage (other methods)

#### **Floating-point Extension**





#### **Floating-point Extension**



- Real (1704) > Int (1662) > Float (1582)
- Constraint solving time of Real (1.7%) < Constraint solving time of Float (10.9%)</li>

## Conclusion



- Input tuning
  - ▶ 10-minute coverage (input tuning) ≥ 1-hour coverage (other methods)
  - SUSY-HMC: 1-hour coverage (input tuning) is about 1.2-2.3X
     higher than 1-hour coverage (other methods)
- Floating-point Extension
  - Floating-point extension using reals achieve 42-122 more branches than the other two



# Thank you!