

# Revisiting the Issue of Performance Enhancement of Discrete Event Simulation Software <sup>1</sup>

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<sup>1</sup>We wish to thank Victor Castillo and the Lawrence Livermore National Laboratory for supporting this research.

This presentation is produced using C. Campani's Beamer  $\text{\LaTeX}$  class.

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*Disclaimer: Our slides here won't show off what Beamer can do. Sorry. :-)*

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- Interpreted languages are slow.
- DES literature mainly algorithm-centric.
- What can be done specifically for interpreted languages?
- What can be done for systems considerations, e.g. VM?

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  - Lots of high-level Python constructs make programming much easier.
  - Python *generator* construct used by SimPy to set up coroutines, i.e. non-preemptive threads.

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- Have class **MachineClass**, with member variables such as **UpTime**, etc.
- Each class has a member function **Run()** which simulates one machine.

## Sample Run() Function

```
def Run(self):
    while 1:
        self.StartUpTime = SimPy.Simulation.now()
        # hold for up time
        UpTime = G.Rnd.expovariate(MachineClass.UpRate)
        yield SimPy.Simulation.hold,self,UpTime
        # update up time total
        MachineClass.TotalUpTime +=
            SimPy.Simulation.now() - self.StartUpTime
        RepairTime = G.Rnd.expovariate(MachineClass.RepairRate)
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The `yield` actually does yield the processor. But `yield` is a coroutine release—next time this function runs, it resumes after the `yield`.

# SimPy Data Structures

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- Assume for simplicity no tied event times.
- The Python list **timestamps** stores all event times, in ascending order. e.g. to determine the earliest scheduled event.  
*A Python list is not an array!* One may insert and delete elements, with the corresponding overhead of shifting data.
- The actual events are in a Python *dictionary* (associative array) named **events**.  
Python dictionaries are implemented as hash tables, reasonably fast.

# SimPy Queue Operations

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Step (i) makes use of Python's **bisect()** function, which performs bisection sort.

That would appear to be  $O(\log n)$  time, for an  $n$ -item event list. Due to SimPy's use of Python's list structure, it is actually  $O(n)$ , due to right-shifting of the data.

# SimPy Dequeue Operations

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When the next event is executed, these operations occur:

- (iii) remove head of list **timestamps**, time  $t$
- (iv) reactivate (invoke Python iterator for) **Run()** function for event of time  $t$  in dictionary **events**

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Again, what would appear to be an  $O(1)$  event is actually  $O(n)$ .



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- Dictionary (smaller problem).
- $O(n)$  insert operation instead of  $O(\log n)$  (big problem).
- $O(n)$  dequeue operation instead of  $O(1)$  (big problem).
- Possible VM issues.

# Our Solutions

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- Rethink event-list algorithms.

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- Needed to overload Python's < operator.

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- SWIG very easy to learn, use.
- We did have to be careful regarding reference counts.

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Testbeds:

- Call center application. Indexed by arrival rates.
- Hold Model. Indexed by coeff. of var. of service times.



# Results

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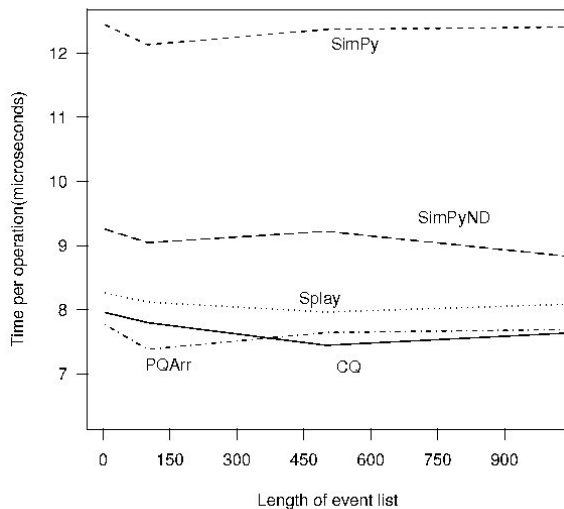
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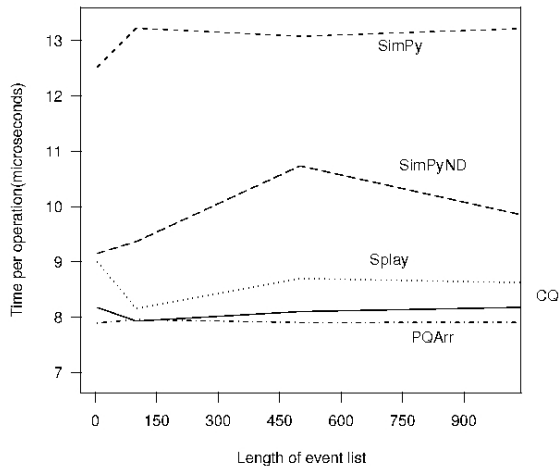
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## Call Center Times Per Op, Lower Traffic

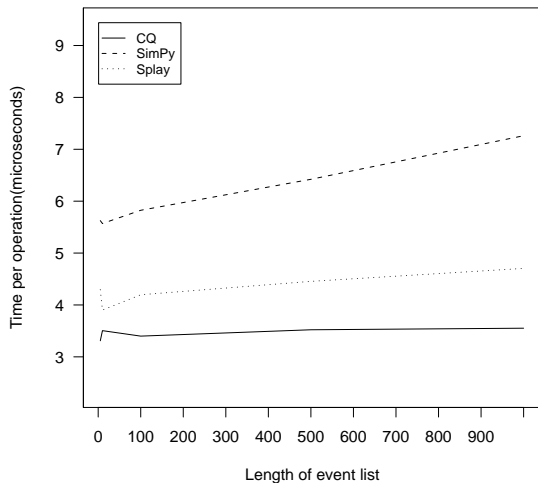


## Call Center Times Per Op, Higher Traffic

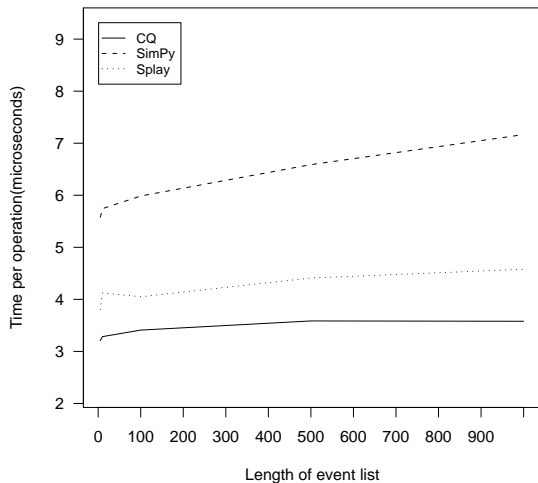




# Hold Model Times Per Op, Smaller COV



# Hold Model Times Per Op, Larger COV



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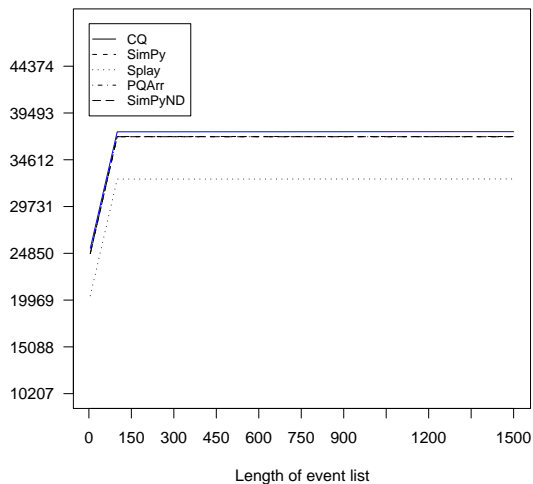
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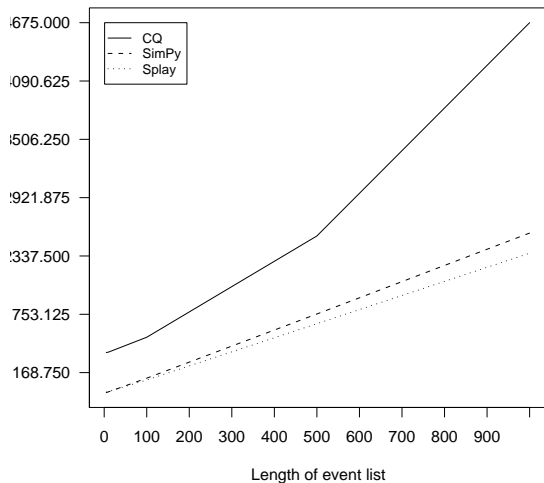
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struct	user time	sys. time	event op. time
PQArr	79.47	4.50	57.87
CQ	33.24	3.95	12.69

# Number of Page Faults, Call Center (lower traffic)



# Number of Page Faults, Hold Model (medium COV)





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- Preliminary experiments on a 64-bit PC, same kernel, suggest greater variability.
- $\therefore$  CQ may do poorly on some systems.

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- Attention to non-algorithmic issues, e.g. paging, may be worthwhile.
- What about JIT? Tried Pyscho but with disappointing results.