Parallel Discrete Event Simulation Course

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Overview
Simulation in General

- **Time**
  - temporal coordinate axis -- simulation time plays a *logical* role
  - program mimics the evolution of the state of a physical system through time

- **Space**
  - there may be a spatial coordinate system as well -- simulation space
  - but even without a coordinate system there is the "space" (namespace) of different simulation objects

- **Spacetime**
  - sometimes it is helpful to think in terms of simulation spacetime

- **State**
  - state of the modeled system that varies in space and time according to rules

- **Event**
  - local change in state at a particular location is space and at a particular time
  - "events" then occupy a single point in spacetime

- **Time and causality obey "Newtonian" semantics**
  - time is global, 1-dimensional, and linear
  - causality only *forward* in time; no causality backward (and restricted sideways) in time
  - no upper bound on distance, velocity, or frequency of interaction
  - simulations must be *deterministic* and repeatable

**What does it mean for space and time to play a logical role**

- It is just as the stack discipline plays a logical role in a procedural program
  - SimTime is global, i.e. Newtonian
  - SimTime is read-only to the model code
- Like the stack pointer in a procedural language, simulation time is not an ordinary variable
  - Stack pointer is controlled by the compiler or interpreter, and the procedural programmer cannot escape them or violate the discipline
  - SimTime controlled by the simulator runtime system (or compiler, interpreter, etc.) but model cannot escape SimTime discipline
- Time and space coordinates are not under the complete control of the simulation programmer
  - They are controlled by the *simulator*, the simulation *platform* code
  - The simulation programmer cannot escape or violate their causal discipline
- time is one dimensional and linearly ordered
  - no branching time or other exotic notion of time
  - sim time never decreases
- sim time must be consistent with Newtonian causality in the model
  - no upper bound on the "velocity" of causal effects, although always finite
  - no causal behavior backward in time
- sim time is virtually always numeric, with sum and difference operations defined
  - this is not actually required for simulation algorithms themselves--linear ordering is all that is required
  - but sums and differences are generally required for simulation models that run on them
- simulation time is *the* synchronization standard used in a PDES (later extended to *virtual time*)
The circled categories are what we will study in this course.
Discrete Event Simulation Applications

- DoD defense & combat models
  - strategic defense models
  - tactical and strategic combat models
- Cyber defense
  - cybersecurity dynamics
- DHS
  - transportation and traffic models
  - regional and national infrastructure models
- Physics
  - particle systems
  - kinetic Monte Carlo models
- Biology
  - epidemiological models
  - population dynamics
  - genetic models
- Digital system
  - network communication — all technologies and protocols
  - telephony
  - digital circuits
  - computer architecture
  - software performance modeling (e.g., transaction systems)
- Mathematics
  - queuing and other systems of stochastic processes
  - birth-death processes
- Others
  - agent models
  - logistical models
  - microeconomic models

What are the objects in the simulation? What are their states? What are the events (state changes)?

Combat models
  objects: units, ships, aircraft, tanks, weapons, communication devices, etc.
  events: moving, sensing, firing, exploding, message transmission, etc.

Transportation:
  objects: vehicles
  events: start on next segment (of road, or flight plan, etc.), arrive at end of segment, interact with another vehicle, crash, etc.

Biology:
  objects: organisms, resources, locations
  events: movement/migration, mutation, birth, infections contact, death, etc.

Network models:
  objects: nodes, routers, switches
  events: packet send, packet arrive, switching decision, routing decision, packet drop, etc.

Economic models:
  objects: humans, corporations (producers, consumers, brokers, banks, markets), gov’t agencies, etc.
  events: financial transactions, demand requests, supply advertisements
History of Discrete Event Simulation

- Origin of discrete event simulation concept and the sequential algorithm for it is hazy
  - Early ideas late 1950s
  - No specific person is credited that I know of.
  - But then, who is credited with the idea of a runtime stack?

- Dozens of programming languages designed specifically for discrete event simulation
  - GASP, SimScript, GPSS, Simula I, Simula 67, ModSim, many others

- Contributions of discrete event simulation to computer science:
  - coroutines
  - object-oriented programming (classes, instances, inheritance)
  - reverse computation (!)

For reference on discrete event simulation programming language history see


Coroutines

- Besides event lists, a key invention introduced in Simula was the notion of a coroutines, a sequential symmetric context switching mechanism (as opposed to subroutine, which is asymmetric and stack-oriented) that could be compiled at user level with no operating system support
- Coroutines were reminiscent of process switching in time sharing systems, but application directed, not time sliced, and with no change in protection domain.
- They are actually closer to mechanically to user level threads, but the distinction between processes and threads had not been made at the time
coroutines each have their own stack
coroutines are not preempted (as threads may be)
when coroutines give up control, it is not to a scheduler, but to another named coroutine (half as many context switches)
It is commonly believed that object-oriented computing originated with Alan Kay and SmallTalk. Not true. While SmallTalk coined the term “object oriented” and made O-O programming popular and more dynamic, the class concept (and the term) originated with Dahl and Nygaard a decade earlier with Simula-67. Furthermore they extended the notion of object-orientedness to include scheduling using simulation time.

Both died in 2002.
The circled categories are what we will study in this course.
Continuous vs. Discrete Simulation
Continuous and discrete simulation have much in common, so this table summarizes the contrasts. There are VERY FEW who are real experts in both continuous and discrete simulation.

Why should we expect unification? Because either can approximate the other to an arbitrary degree of precision.

Mixed discrete and continuous systems:
fairly straightforward affair for mixed discrete and ODE-derived systems, but no one knows how to unify discrete and PDE-derived systems

Grand challenge: fully unify the computational methods for discrete models and multidimensional continuous models.

I have long felt that the world could benefit from a unification of the two, but so far no one has stood up to accomplish it. If anyone here is deeply knowledgeable about continuous simulation and wants to collaborate with me in a unification, let me know!
Discretization of a continuous model is the transformation from equational form into a discrete model. Virtually always this is a time-stepped model.

Grand challenge: Develop techniques for transforming continuous models into discrete event models. No one knows how to do that in general, although there are hints coming. Would solve the general problem of multiscale simulation, in time at least.
Time-stepped vs. Event-driven
Discrete time-stepped vs. discrete event models

• The fundamental difference is **not**:
  • integer time vs. continuous (floating point) time
  • discrete state space vs. continuous (floating point) state space
  • synchronous vs. asynchronous

• The fundamental difference is:
  • time-stepped models: event times and locations are **decided statically** (or quasi-statically)
  • event-driven models: event times and locations are **dynamically computed**
    • no restrictions regarding locality or scale in time or space
    • one consequence: one-sided messaging preferred to two-sided

difference is analogous to other static vs. dynamic paradigms
  arrays vs. strings or lists
  static storage vs. stack or heap storage

With (quasi)static knowledge of the simtimes and communication patterns built-in to the code, it is possible for the receiving side of the code to know when and how many messages to expect at a given point in the logic. Hence 2-sided messaging with explicit RECEIVE() primitives are appropriate.

But in PDES the simulator has no static knowledge of the communication pattern or the times at which an object will receive an event message, it is impossible to know if or when a event message will arrive, or how many, so an explicit RECEIVE primitive is and any attempt to use two-sided message primitives requires systematic use of polling.
## Time-stepped vs. Event-driven Simulation

<table>
<thead>
<tr>
<th></th>
<th><strong>Time-stepped</strong></th>
<th><strong>Event-driven</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td>Event Times</td>
<td>statically chosen</td>
<td>dynamically computed</td>
</tr>
<tr>
<td>Communication Patterns</td>
<td>mostly static</td>
<td>dynamically computed</td>
</tr>
<tr>
<td>Simple Implementation</td>
<td></td>
<td>more complex implementation</td>
</tr>
<tr>
<td>Static Lower Limit on Time Scale</td>
<td></td>
<td>no lower limit on time scale; inherently multiscale</td>
</tr>
<tr>
<td>Events in Spacetime</td>
<td>dense and regular</td>
<td>sparse</td>
</tr>
<tr>
<td>Clumsy and Inefficient to Couple Two Time-stepped Simulations</td>
<td></td>
<td>easy and natural to couple two event-driven simulations</td>
</tr>
<tr>
<td>Parallel Models</td>
<td>generally SPMD with 2-sided comm primitives</td>
<td>generally MPMD with 1-sided comm primitives</td>
</tr>
<tr>
<td>Appropriate</td>
<td>for spatially and temporally regular models</td>
<td>for irregular, asynchronous models</td>
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**Time-stepped**
- A key consequence is that there is a static limit to how fine a time scale is used in a simulation.
- Continuous simulations are discretized by transforming the continuum model (equation) into a discrete time-stepped simulation.
  - AMR is an attempt to get around the limitations of that approach.
- Parallel time-stepped simulations need not, but usually do, run the exact same algorithm at every spatial point and are SPMD.

**Event-driven**
- Event-driven simulations the times of events are calculated dynamically.
- Also generally the event locations are computed dynamically; thus the communication pattern can be arbitrarily irregular.
- In such a case there is no limit to how fine a time scale may evolve in a simulation.
- In this respect, event-driven simulation is more general than time-stepped simulation as a paradigm.
  - Imagine how continuous simulation might change if equations were discretized as event-driven rather than time-driven models!
  - AMR is a step in this direction, for the case when there is a spatial continuum as well.
  - But no one has a full theory yet about how to do this right.
- MPMD
  - Event-driven simulations usually have more than one type of object, and are MPMD.
Sequential Discrete Event Simulation
**Sequential Discrete Event Simulation**

- **simTime** -- global real or integer (assume real)
  - global simulation time; read-only to model code

- **objects** -- class instances (e.g. “objects” in OO sense)
  - represent discrete objects in the system being simulated
  - also called LPs (logical processes)

- **state** -- the field values of an object
  - represents the physical state of the object being modeled

- **event** -- execution of a method on an object at a simTime
  - represents a discontinuous change in one object’s state

- **eventNotice** -- event (method call) scheduled for a future time
  - represents an event that will (or might) happen in the future
  - stored until the appropriate simulation time

- **eventList** -- global priority queue of eventNotices
  - ordered by simTime
  - represents the currently known scheduled future events
  - events insert eventNotices into eventList (and in some paradigms, delete them)

The event list is analogous to the runtime stack of procedural programming language. It is a priority queue instead of a stack, but like the stack it organizes the many method/event calls that make up the computation.
4 objects, A, B, C and D with their internal states

current simTime is 19.4

event list has 5 events scheduled in it

simTime is always the same as the lowest simTime event in the event list

event list is not really implemented as a linear list—it is a priority queue, generally a tree, i.e.
a heap, red-black tree, or (ideally) a splay tree

A is executing the move method at time 19.4
**Sequential DES algorithm**

```plaintext
createInitialObjects();
eventList.insert(initialEvents);

while (! ( terminationCondition() || eventList.empty() ) ) do {
    event e = eventList.removeMinSimTime();  // Choose next event
    simTime = e.getEventTime();  // set simTime and unpack event
    object = e.getEventObject();
    method = e.getMethod();
    args = e.getArgs();

    object.method(args);    // May change state of object
    // May insert future events into eventList
    // May create or destroy objects
    // May delete future event from eventList
}

finalize();
```

This is an object-oriented style of DES

We distinguish the simulator code from the simulation code

- Blue text is the simulation code--the model itself
- Black text is the simulation algorithm itself--the platform

Besides initialization and termination issues, all of the work of the model code is in the object methods

Assume no ties and no zero-delay in simTime

- an event at time $T$ can only schedule events at times strictly $>T$
- no two events for the same object have the same simTime
  - but two events for different objects may have the same time
  - hence simTime strictly nondecreasing with every event

Reasons for, refinements of, and consequences of these assumptions will come later

- surprisingly complex, especially for parallel sim
- must prepare the ground
Sequential Discrete Event Simulation

simTime: 19.4

eventList:

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</thead>
<tbody>
<tr>
<td>A</td>
<td>B</td>
<td>A</td>
<td>C</td>
<td>D</td>
</tr>
<tr>
<td>move</td>
<td>turn</td>
<td>fire</td>
<td>observe</td>
<td>observe</td>
</tr>
<tr>
<td>(3.2,4.5)</td>
<td>(L,45)</td>
<td>()</td>
<td>(3.2,12.4)</td>
<td>(-2.1,-6.8)</td>
</tr>
</tbody>
</table>

objects:

A: [ ] [ ] [ ] [ ]
B: [ ] [ ] [ ] [ ] [ ]
C: [ ] [ ] [ ]
D: [ ] [ ] [ ]

4 objects, A, B, C and D with their internal states

current simTime is 19.4

event list has 5 events scheduled in it

simTime is always the same as the lowest simTime event in the event list

event list is not really implemented as a linear list—it is generally a tree, i.e.
    a heap, red-black tree, or (ideally) splay tree

A is executing the move method
The move method inserted no events into the eventList, so the next event is at time 19.6 for B

Note A's state has changed
B's state has changed, but note that B inserted two events into the event list.

One it happened to be for itself, and happened to be the very next event.

It also inserted an event for D

Hence and event for B runs next also.
B’s event caused a new object, E, to be created and also caused an event for E to be scheduled.

That’s all there is to the core sequential algorithm! The main variations in this algorithm at the level we are talking about are in the data structures used for the event list!

But efficiently parallelizing this algorithm is a very complex undertaking and requires a lot of parallel computation issues to be rethought from the ground up, as we will see when we get to optimistic parallel algorithms.
That’s all there is to the sequential DES algorithm!

• The most significant variations are in the implementation of the priority queue for the event list.

• This algorithm is optimal in both time and space!

Efficient, scalable parallelization of this one simple algorithm is what the rest of the course is about.