Tutorial at AICCSA'07 (May 13-16, 2007, Amman, Jordan)

Title: Introduction to Accurate Stochastic Discrete-Event Simulation

Duration:

a half day tutorial

Instructor:

Krzysztof (Krys) Pawlikowski Professor in the Department of Computer Science and Software Engineering University of Canterbury in Christchurch, New Zealand

Summary:

The computer revolution initiated in the twentieth century has resulted in the adoption of computer simulation as the most popular paradigm of scientific and engineering investigations. It has become the most commonly used tool in performance evaluation studies of such complex dynamic stochastic systems as telecommunication networks. Such reliance on simulation studies raises the question of credibility of the results they yield.

In this tutorial, we will discuss **main problems and solutions of quantitative stochastic discrete-event simulation**, i.e. the stochastic simulation in which the emphasis is put on statistical correctness of the final results. Whole spectrum of the problems will be covered: from generators of pseudo-random numbers to methods of controlling precision of the final results in sequential stochastic simulation conducted in a distributed mode on multiple computers of a LAN. The latter will be discussed in the context of Multiple Replications in Parallel (MRIP) scenario, which allows to speed up stochastic simulation by launching multiple simulation engines cooperating in production of data for central analysers (one central analyser for each performance measure). An implementation of MRIP in a simulation package Akaroa2, a universal controller of distributed quantitative simulation, which automatically launches simulations on an arbitrary number of simulation engines and controls precision of an arbitrary number of analysed performance measures, both in terminating and steadystate simulation, will be also discussed.

The tutorial will conclude with a survey of open research problems of quantitative stochastic simulation

Outline of Topics:

The topics will be discussed in the following order:

- Introduction to quantitative stochastic simulation: basic terms and definitions. Necessary conditions of a trustworthy simulation study.
- Sources of randomness: theory and practise of modern generators of pseudorandom numbers for simulation on single and multiple processors.
- Sequential quantitative stochastic simulation: its principles and implementation in terminating, steady-state, and non-stationary simulation; measures of precision of the final results.
- Automation of precision control in terminating and steady-state simulation: survey of methods of analysis. The initial transient period in steady-state simulation: theory and methods for estimating its length.

• Multiple Replications in Parallel (MRIP) for speeding up quantitative stochastic simulation. Applications of MRIP in Akaroa.2.

Additional readings:

- M. Law and W. D. Kelton. *Simulation Modeling and Analysis*. McGraw-Hill, 2006.
- K. Pawlikowski. Steady-state Simulation of Queueing Processes: A Survey of Problems and Solutions. *ACM Computing Surveys*, June 1990, 123-170

Prerequisities.

It is assumed that participants will be familiar only with the very basic concepts of probability theory and statistics.

Biography of the instructor:

Speaker Biography

Prof. Krzysztof (Krys) Pawlikowski has over 30 years experience as a university lecturer. Currently he is a Professor in Computer Science and Engineering at the University of Canterbury, in Christchurch, New Zealand.

He received his PhD degree in Computer Engineering from the Technical University of Gdansk, Poland. The author of over 140 research papers and four books, has given invited lectures at international conferences and at over 80 universities and research institutes in Asia, Australia, Europe and North America.

His research interests include performance modelling of multimedia telecommunication networks, tele-traffic modelling, computer simulation and distributed processing.

Target Audience:

Research students, scientists and engineers who use discrete-event simulation in performance evaluation studies of computer systems and networks, or any other stochastic dynamic systems, such as communication or inventory systems, production lines, etc.

Introduction to Accurate Stochastic Discrete-Event Simulation

Krzysztof Pawlikowski

University of Canterbury Christchurch



K. Pawlikowski











University of Canterbury:

Interdepartmental

Network Research Laboratory:

Protocols, Distributed Processing

and Simulation

(1) Simulation Research Group:

developing new methodology of distributed and automated stochastic discrete-event simulation

(2) Network Research Group:

modelling and evaluating the Internet and other multimedia networks in wired, wireless and optical technologies



Membership:

- Prof. Dr K. Pawlikowski (CSSE)
- Prof. Dr H. Sirisena (ECE)
- Associate Prof. Dr D. McNickle (MGMT)
- Dr. Greg Ewing (research associate)
- Master and PhD students of CSSE, ECE and Management

In 2000-2005:

5 PhD graduates (3 from CSSE) over 10 Master graduates (4 from CSSE)

Currently: 4 PhD and 6 Master students (1 PhD and 2 MSc students of CSSE)

see www.cosc.canterbury.ac.nz/research/RG/net_sim/



Computer simulation:

one of the most important factors enabling new developments in science & technology

Congress of the USA, 1996

Three paradigms of science:

- theory
- experimentation
- computer simulation



Interdisciplinary character of Computer Simulation



Current controversy: Should **Computer Simulation** be considered as a new discipline (of Computing) ? See discussion within the Society for Modelling and Computer Simulation.

Departments of (Modeling and) Computer Simulation have

been established at a number of universities world-wide ...

Quantitative stochastic simulation:

used in performance evaluation studies of dynamic stochastic systems and processes

In particular, studies of

Telecommunication Networks, motivated by:

- Enormous dependence of our civilisation on computers and their networks makes their performance evaluation mandatory before they are deployed.
- Modern multi-media networks have become so complex that their performance can be usually studied only by simulation.

Advanced Research Project Agency of US Department of Defense lists **Network Modelling & Simulation** among the most important 23 areas of Information Technology which they support.



"DARPA Information Processing Technology Office will create a new generation of computational and information systems that possess capabilities far beyond those of current systems".

For this purpose 23 IT programmes have been identified, including:

Architectures for Cognitive Information Processing **Bio-Computation Compact Aids for Speech Translation Coordination Decision Support Assistants** Effective, Affordable, Reusable Speech-to-Text Fast Connectivity for Coalitions and Agents Project High Productivity Computing Systems Learning Applied to Ground Robots Mobile Autonomous Robot Software **Network Modeling and Simulation** Organically Assured and Survivable Information Systems Personalized Assistant that Learns **Polymorphous Computing Architectures** Power Aware Computing and Communication Quantum Information Science and Technology **Real-World Reasoning** Self-Regenerative Systems Software for Distributed Robotics Taskable Agent Software Kit

see *www.darpa.mil/ipto/index.htm* (in July 2005)



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Challenge:

to find solutions (protocols, architectures, topologies) which can ensure the best (or the worst) performance of networks



Issue of credibility of simulation studies:

- When can results obtained by quantitative stochastic simulation be regarded as **accurate** ?
- How to ensure credibility of quantitative analysis based on stochastic simulation ?



General guidelines for conducting valid performance evaluation studies based on stochastic simulation:

- use a valid simulation model:
 - correct functional representation of simulated mechanisms
 - appropriate stochastic characteristics of the model
- execute valid simulation experiment:
 - use appropriate pseudo-random number generators (PRNGs)
 - apply appropriate analysis of simulation output data



SOURCE(S) OF PRIMARY RANDOMNESS:

Good PRNGs (of uniformly and independently distributed numbers):

- should generate numbers that satisfy the most rigorous statistical tests (of uniformity and independency),
- should be able to generate multiple streams of numbers in long cycles,
- should be fast and accurate.



Until now, the most popular PRNGs are linear congruential generators with cycle $L \le 2^{48} - 1$ (with $L \le 2^{31} - 1$ being still the most popular)

Taking into account recent developments in science and technology, one can show these "baby" generators cannot be considered as good PRNGs and should be replaced by much more powerful ones.



Cycle length vs. computing technology

Proposition:

The length of a typical simulation (measured by CPU time) does not become shorter as computing technology becomes more powerful: **users use faster computers for simulating more complicated processes within the available time**.

Conclusion 1:

PRNGs with adequately long cycles are needed to avoid repetitions of Pseudo-Random Numbers (PRNs) during a typical simulation.



Technology & generation rate of PRNs

Moore's law by Gordon Moore (1965) see research.microsoft.com/rgray/Moore_Law.html

Law of accelerating returns by Ray Kurzweil (2001) see *www.kurzweilai.net/articles/art0134.html*

On average, computing power of computers doubles each 1.5 - 2 years

At University of Canterbury in Christchurch, New Zealand:

In 1995, workstations with CPU clock at 150 MHz using rand (from C library): 10^6 PRNs in about 2 seconds, or 2^{31}_{48} PRNs in about 72 minutes 2^{48}_{48} PRNs in about 18 years

In 2000, PCs with the CPU clock at 800 MHz using rand (from C library): 10⁶ PRNs in about 0.4 seconds, or

 2^{31} PRNs in about 13.5 minutes 2^{48} PRNs in about 3.5 years

 $\frac{\text{In 2002, PCs with the CPU clock at 2.4 GHz}}{\text{using rand (from C library): } 10^6 \text{ PRNs in 0.13 seconds, or}}{2^{31} \text{ PRNs in about 4.5 minutes}}$

2⁴⁸ PRNs in less than 1.1 years





Effect of Moore's Law on Processor Speed



Cycle length of PRNGs vs. theory

Birthday Spacing Problem:

one of reference problems/tests for testing uniformity of PRNGs see D.E.Knuth. Art of Computer Programming, vol.2

Recent finding:

Any linear congruential PRNG fails the Birthday Spacing Test if one applies this test to $n \ge 16 \sqrt[3]{L}$ numbers generated by that PRNG, where L is the length of its cycle.

L'Ecuyer and Simard; in Mathematics and Computers in Simulation, 2001.

Conclusion 2:

During a simulation, a PRNG should be used as a source of **no more than** $16\sqrt[3]{L}$ **numbers** (linear congruential PRNG), or **no more than L numbers** (non-linear congruential PRNG).



Using popular linear congruential PRNGs on computers with 2.4 GHz CPU clock

Cycle length	n_{max}	Max. total time of "stochastically safe" application of a good PRNG
$L = 2^{31}$	20 594	less than 0.3 seconds
$L = 2^{48}$	1 048 576	less than 14 seconds

where $n_{max} = 16 \sqrt[3]{L}$



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Search for good PRNGs of the 21st century

M. Schoo, K. Pawlikowski & D. McNickle, Technical Report TR-COSC 03/05

Assumption:

the process of PRN generation takes no longer than 1% of simulation time.





Required cycle length for linear generator on multiple processors

Distributed simulation on 2^{14} processors, with 100 THz CPUs, needs a linear congruential PRNG with the cycle length of about 2^{160} if it lasts 1 hour, and the cycle length of about 2^{182} if it lasts 1 day.





Required cycle length for non-linear generator on multiple processors

Distributed simulation on 2^{14} processors, with 100 THz CPUs, needs a nonlinear congruential PRNG with the cycle length of about 2^{57} if it lasts 1 hour, and the cycle length of about 2^{65} if it lasts 1 day.



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Conclusion:

"Statistically safe" PRNGs of the 21^{st} century should be able to generate numbers in cycles of length $L > 2^{180}$ (linear PRNGs), or $L > 2^{65}$ (non-linear PRNGs).



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Multiple Recursive Congruential PRNGs:

G1:

$$\begin{aligned}
r_i &= (2109532706r_{i-1} + 1651737654r_{i-7}) \\
\mod (2^{31} - 19) \\
L &= (2^{31} - 19)^7 - 1 \approx 2^{217}
\end{aligned}$$

Multiple streams of PRNs generated by **Cycle Splitting**. Proven good uniformity in up to 20 dimensions.

For other properties and portable implementations, see: *L'Ecuyer, Blouin and Couture. ACM Trans. on Modelling and Computer Simulation, April 1993.*



Combined Multiple Recursive Congruential PRNGs:

G2:

$$\begin{aligned} r_{1,i} &= (1403580r_{1,i-2} - 810728r_{1,i-3}) \\ & \mod (2^{32} - 209) \\ r_{2,i} &= (527612r_{2,i-1} - 1370589r_{2,i-3}) \\ & \mod (2^{32} - 22853) \\ r_i &= (r_{1,i} - r_{2,i}) \mod (2^{32} - 209) \\ L &\approx 2^{191} \end{aligned}$$

Multiple streams of PRNs generated by **Cycle Splitting**. Proven good uniformity in up to 45 dimensions.

For other properties and portable implementation of G2, see: *P. L'Ecuyer. Operations Research, 1996.*



Generalised Feedback Shift Register (GFSR) PRNGs:

Twisted GFSR PRNG known as Mersenne Twister (MT19937) with the cycle $L = 2^{19937} - 1$

M. Matsumoto and T. Nishimura. ACM Trans. Mod-

elling and Computer Simulation, 1998

Some properties of MT19937:

- uniformity in up to 623 dimensions;
- multiple streams of PRNs generated by dynamic creation of multiple Mersenne Twisters

Warning: There is a problem with initialization. Special care has to be taken when selecting the seed, as a long sequence of strongly correlated numbers can be generated !



Note:

Non-linear PRNGs (Inversive Congruential and Explicite Inversive Congruential) are currently too slow to be practical in real applications. A fast algorithm for modular inversion is needed.

Our recommendation:

- Use one of linear congruential PRNGs proposed by Pierre L'Ecuyer.
- Multiplicative Lagged-Fibonacci PRNGs can be also considered.



SIMULATION OUTPUT DATA ANALYSIS:

ata ah a ati a		simulated
stochastic	≡	statistical
simulation		experiment

Results of any quantitative simulation study can be credible only if they are obtained with an acceptable (low) statistical errors.

Otherwise ... the results can be meaningless !



Which of these two sets of results are more conclusive/credible ?



Analysis of TCP: results of one simulation vs means over 100 simulations





Evaluation of a Medium Access Control protocol of a mobile communication network. Results with relative statistical errors not greater than (a) 25%, (b) 1%, at 0.95 confidence lavel.



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The only practical solution for producing credible results from simulation: sequential on-line data analysis of statistical errors at consecutive checkpoints of simulation



the length of simulation is determined

during simulation:

the simulation is continued until statistical errors of results become satisfactorily small



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Are simulation results reported in scientific literature appropriately analysed ?



Results of a survey

of 2246 publications on networks

published between 1992-1998 in:

- (i) Proc. of IEEE INFOCOM
- (ii) IEEE Trans. Communications
- (iii) ACM/IEEE Trans. Networking
- (iv) Performance Evaluation Journal

show that,

over 51% papers reported simulation results

see [Pawlikowski et al., IEEE Comms., Jan. 2002.]





Survey of 2246 papers on networks from

More than 51% of all papers reported results obtained by simulation

Pawlikowski et al. IEEE Comms., Jan. 2002



Proportion of papers reporting statistically analysed simulation results.



77% of papers that reported results based on simulation gave no evidence at all that simulation output data were statistically analysed.

Pawlikowski et al., IEEE Comms., Jan. 2002.





Proportion of all papers on stochastic simulation that reported results analysed statistically:

from Terminating Simulation (**TS**), and from Steady-State Simulation (**SS**). **NN** = No information about the type of simulation nor about analysis of results



Results of this survey:

77% of papers that reported results based on simulation either

- gave no details of the simulation, or
- did not inform on how data were analysed (were they analysed at all ?), or
- both.

While all such papers reported non-repeatable simulations, some of them reported purely random results only !



Another survey:

Survey of all **151 papers published in Proceedings of the ACM Int. Symposium on Mobile Ad Hoc Networking and Computing** in 2000-2005:

- 75.5% of the papers reported results from simulation
- less than 15% of the papers reported repeatable studies of networks
- 12% of papers reported results which were based on sound statisitcal analysis of simulation output data.

See: S. Kurowski, T. Camp and M. Colagrosso. "MANET Simulation Studies: The Incredibles".



Simulation studies of telecommunication networks:

- PSEUDO-SCIENCE ? (results based on personal beliefs of their authors)
- SCIENCE ? (results obtained and reported according to the principles of the scientific method)

... any new scientific hypothesis should be at least independently testable ...

Karl Popper (1901-1994), philosopher of science



Guidelines for presenting a credible publication based on simulation results:

- make sure that your simulation program represents valid simulation model;
- use well tested PRNG with sufficiently long cycle
- describe your statistical analysis of simulation output data in the paper, or refer to a technical report



Resolution to the credibility crisis of quantitative simulation:

- professional certification of simulation specialists ?
- automated sequential simulation ?

See *www.simprofessional.org* for discussion on professional certification of simulation specialists, proposed by Society for Computer Modelling and Simulation Int.



Problems of quantitative simulation studies of telecommunication networks:

- limitations of statistics: limited spectrum of methods for sequential analysis of simulation output data
- technological limitations: prohibitively long simulation time can be needed for securing sufficiently low statistical errors

For example, steady-state mean delay in $M/M/1/\infty$ buffer, with 5% error, at 0.95 confidence level, needs:

(a) **over 61 000 000** observations ($\rho = 0.99$),

(b) over 6 143 000 000 observations

if $\rho = 0.999$.



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Possible solution:

application of distributed processing (distributed generation of simulation output data)



Multiple Replications in Parallel (MRIP):

Multiple simulation engines independently produce statistical data by running their own simulation replications, but cooperate together in data analysis.



Main properties:

- model-independent parallelisation
- no parallel programming needed
- possible automation of parallelisation



MRIP as a (distributed) simulation in multiple time streams:





Truncated Amdahl's law of speedup in MRIP



assumption: 1000 checkpoints needed f = relative length of non-parallelisable stage of simulation (for example: the length of initial transient phase in steady-state simulation)

See K. Pawlikowski and D. McNickle. "Speeding Up Stochastic Discrete-Event Simulation". Proc. ESS'2001, Marseille, October 2001.



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IMPLEMENTATION OF MRIP

AKAROA 2 = pAcKage for Automatic geneRation and process cOntrol of pArallel stochastic simulation (version 2)

written in C++

tested on Sparc/SunOS4 & Solaris2, i386/Solaris2 and Linux

automatic parallelisation

automatic data collection and on-line analysis



First reported automated implementations of MRIP:

EcliPSe [Rego & Sunderam, 1991]

AKAROA [Pawlikowski & Yau, 1992]



Example:



In the system loaded at 99%, analysis of the mean processing time of a job,

with error not exceeding 5%,

at 0.95 confidence level, requries collection of about 50 000 000 observations.



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```
/* * mm1.C - M/M/1 Queueing System */
#include "akaroa.H"
#include "akaroa/distributions.H"
#include "akaroa/process.H"
#include "akaroa/resource.H"
double arrival_rate; // Rate at which customers arrive
double service_rate; // Rate at which customers are served
// Server, modeled as a Resource with a capacity of 1 unit.
Resource server(1);
class Customer : public Process {
public:
 void LifeCycle();
};
void Customer::LifeCycle() {
  Time arrival_time, time_in_system;
  arrival_time = CurrentTime();
  server.Acquire(1);
  Hold(Exponential(1/service_rate));
  server.Release(1);
  time_in_system = CurrentTime() - arrival_time;
  AkObservation(time_in_system);
}
// The main program. After getting the load from the
// command line and calculating the arrival and service
// rates, we enter a loop generating new customers
// at the arrival rate.
int main(int argc, char *argv[]) {
  real load = atof(argv[1]);
  service rate = 10.0;
  arrival_rate = load * service_rate;
  for (;;) {
    new Customer;
   Hold(Exponential(1/arrival_rate));
```



Main functional blocks of AKAROA.2

akrun (shell command interface)

- initiates simulation
- reports results to user

akmaster (Parallel Simulation Manager)

- manages launching of simulation engines
- receives checkpoint data and calculates global estimates
- controls simulation run length

akgui: AKAROA.2 graphical user interface

Interprocess Communication based on UNIX Internet-domain stream sockets



BLOCK DIAGRAM OF AKAROA.2







AKAROA.2 architecture

(2 simulations in progress, each one with3 engines run on separate network hosts).



Akgui:

Akaroa

	Hosts					Nou Simulation	
No	D.	Hostname			PID	No. Engs	New Simulation
	1	s445			10776	0	0
	2	s452			1033	0	Guit
	3	s448 - 450			13746	U	
	4	s450			9998 1050	0	
	6	s442 s420			1200	0	
	7	3439 =444			3408	ů –	
	8	s438			2087	Ō	
	9	s432			322	0	
	10	s433			2091	0	
				Sin	nulations		
				-			
SID)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand
SIC)	No. Params	No. Engs			Com	mand



Akgui: before launching new simulation

New Simulation					
Command and argu	iments:				
mm1 0.1					
Precision	0.05				
Confidence	0.95				
AnalysisMethod	Spectral				
CPSpacingFactor	1.5				
CPSpacingMethod	Linear				
No. of engines:		3			
💊 Select Hosts					
			Run	Cau	ncel



Akgui: during simulation

mm10.9								
Command: mm1 0.9 Simulation ID: 6					Add Engines			
Status:	Status: Running						Close Window	
	Simulation Engines							
EID		ŀ	lost	PID	State	1		
1 2 3	s445 s452 s448			26990 22934 2635	alive alive alive			
Param	Param Relative Precision							
1	1 1e-2 2e-2 3e-2 5e-2 1e-1 2e-1 3e-1 5e-1							
—	Global Estimates							
Param	n Mean Precision Total Obs Trans Obs		Trans Obs	Chkpts/Min	Last Chkpt Arrived			
1		1.02561	0.2570	62286	2838	26	Thu May 27 14:35:57 1999	



Akgui: the end of simulation

mm10.9 mm1 0.9 Command: Add Engines... Simulation ID: 6 Finished Status: **Close Window** Results **Kill Simulation** Total Obs Trans Obs Param Mean Delta +/-1.0155 0.0496277 396813 2838 1



Some special features of AKAROA2:

- dynamic parallelisation
- fault tolerance (*if simulation engine is lost, simulation continues with remaining engines*)



AKAROA 2 as MRIP controller of other simulation packages

- AKAROA2/Ptolemy interface (1997, Ewing and Pawlikowski, University of Canterbury, New Zealand)
- AKAROA2/NS2 interface (2000, TU Berlin, Telecommunication Networks Group)
- AKAROA2/OMNET++ interface (2002, TU Berlin, Telecommunication Networks Group)



Akaroa 2 is offered as a free-ware for non-profit research at universities.

See www.cosc.canterbury.ac.nz/~krys.

A counter on Akaroa2 webpage has recently showed about 11 000 visits and about 800 downloads of the code (during last 5 years)



Akaroa2 is used at universities in over 60 countries

(Jan. 2006), including: Australia Belarus Belgium Bosnia Brazil Canada Chile China Colombia Cyprus Denmark England Finland France Germany Greece Guana Hong Kong Hungary India Indonesia Iran Ireland Italy Jakarta Japan Jordan

Korea Lithuania Macedonia Malaysia Mexico New Zealand Nigeria Norway Pakistan Poland Russia Singapore Slovakia South Africa Spain Switzerland Taiwan Thailand The Netherlands Trinidad Tunisia Turkey Ukraine United Kingdom United States Vietnam

Commercial users include IBM Research Labs in Zurich, Switzerland.



Currently:

Akaroa2 can automatically produce accurate estimates of mean values, probabilities and proportions.

Work in progress:

sequential analysis of quantiles and probability distributions.



Future work:

- sequential analysis of rare events
- MRIP of distributed simulation models (distributed simulation on multiple clusters of workstations)
- distributed optimization in MRIP scenario.

