Effective Monte Carlo simulation on System-V massively parallel associative string processing architecture

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Abstract

We show that the latest version of massively parallel processing associative string processing architecture (System-V) is applicable for fast Monte Carlo simulation if an effective on-processor random number generator is implemented. Our lagged Fibonacci generator can produce $10^8$ random numbers on a processor string of 12K PE-s. The time dependent Monte Carlo algorithm of the one-dimensional non-equilibrium kinetic Ising model performs 80 faster than the corresponding serial algorithm on a 300 MHz UltraSparc.
1 Introduction

Massively parallelism appears nowadays mainly on the level of MIMD processor clusters owing to the commercially available cheap building elements with ever increasing clock speeds. However everybody knows that CPU clock speed can not increased without limit, and the memory access speeds are well behind. Therefore redesigning of the high performing architectures are necessary time to time. One such a direction is the intelligent memory (IRAM [1]) or processor on memory projects. By putting more and more fast memory on the silicon surface of the processors (cashes) or processors at the edges of the memory matrices one can avoid huge (1000 times magnitude) losses on the connection hardware, buses.

The Massively Parallel Processing Collaboration [2] started a research and development of conceptually similar architectures in the early nineties with a target of processing large quantities of parallel data on-line. The basic architecture was a low level MIMD high level SIMD to fit the best to the requirements. While the development has stopped with prototype (ASTRA-2) in the physics research development collaboration, the founding engineering company ASPEX continued developing the ASP architecture to produce a special “co-processor” for workstations that enhances image processing capabilities.

The latest architecture System-V has already proven its image processing power [3]. We demonstrate is this work that it is also applicable for Monte Carlo simulations in statistical physics. In section 2 we introduce the basics of the hardware of the new architecture, while in sections 3 and 6 we show how effective random generation and simulations can be coded. In section 5 we introduce the statistical physical models and the time dependent algorithms to measure critical exponents. More detailed analysis of the results will be discussed elsewhere [15, 20].

2 The System-V architecture

System-V is a specific VMEbus system implementation of the Modular MPC (Massively Parallel Computing) architecture. It provides programmable, cost-effective yet flexible solutions for high performance data-intensive computing applications, such as signal and image processing and 2D and 3D image generation problems. The architecture of System-v is modular, so that the configuration of processors, memory, data channels and I/O units in a system can be independently scaled and exactly balanced to give the most economical and effective application solution. Application development is achieved by programming in C.

System-V derives its computational flexibility and high-performance from a programmable SIMD architecture, namely the ASP (Associative String Processor) [4]. In addition to the SIMD core, high performance RISC processors and custom ASICs are used to issue program instruction and manage data I/O.

![System-V architecture schematic diagram](image)

As indicated in Figure 1, the key features of the System-V architecture are multiple data channels with overlapped data transfer and processing, independent scalability of processors and memory in each channel, and multiple external I/O paths. The major building block of the architecture is the Secondary Data Transfer (SDT) Channel, comprising the ASP, Secondary Data Movement Controller (SDMC) and Secondary Data Store (SDS). SDT channels can be replicated to increase SDT bandwidth. Each channel also has its own high-speed Tertiary Data Interface (TDI) for direct high-speed external I/O.

As described in a following sub-section, the ASP contains an array of Associative Processing Elements (APEs) and a Primary Data Store (PDS). Each APE has a private Primary Data Transfer (PDT) channel to the PDS. In addition, all APEs are connected via a flexible communications network.

The PDS is connected to the SDS by the SDMC, which performs secondary data transfers between the two. Its also controls access to the SDS for Tertiary Data Transfers (TDT) from external devices or the Main Processor.

The Low-level ASP Controller (uLAC) performs global control of the ASPs. The Main Processor can be a conventional workstation, PC, CPU card or micro-processor. To control System-V it must perform the tasks of Instruction Stream Management (ISM) and Data Stream Management (DSM). It does this by issuing commands to the uLAC and the SDMCs.
As depicted in Figure 2, comprises a VMEbus motherboard on which sits a stack of daughter-boards with the option of smaller mezzanine cards installed on-top of the daughter-boards.

In the heart of System-V flexibility and scalability is its stacking buses. Three high performance buses called AGbus, SIBus and DSMbus run through all the cards in the System-V stack. A further two nested buses provide APE and memory expansion on the SDT Channel cards.

The ASP Global Bus (AGbus), a synchronous bus, carries instructions from the instruction stream manager on the stack base card to the ASP devices on the APE array daughter-boards.

A single System Interface bus (SIBus) connects all the cards in the System-V stack to the management processors on the base card and to the VMEbus.

The Data Stream Manager bus (DSMbus) connects the data stream management processor on the base card or on a daughter-board to all the cards in the System-V stack between it and the next DSM card above it. There can be up to eight DSMbuses in a single System-V stack, all of which can operate in parallel.

The Stackable APE Module bus (SAMbus) allows the number of APEs to be scaled per SDT channel by plugging in more SAMs.

Finally, the memory expansion bus allows the amount of memory per SDT channel to be scaled by plugging in Stackable Memory Modules (SMM)s.

Multiple System-V SIMD stacks can be connected together in pipeline or processor farm topologies to provide even higher levels of performance.

2.1 System-V building blocks

The hardware blocks that can be used in building a System-V configuration are 4 cards. The base or ISM card combines the ISM and DSM functionalities of System-V. It is a VMEbus card that provides the interface between the SIMD stack and the rest of the VME system. It features two SPARC processors and a uLAC-1001 co-processor for DSM and ISM functions respectively.

The APE Array (APEA) card is SIMD stack daughter-boards that implement a Secondary Data channel containing an Associative Processing Element (APE) Array, Secondary Data Store (SDS) and a Secondary Data Movement Controller. The number of APEs and size of memory can be independently scaled by SAM and SMM mezzanine cards.

The Stackable APE Module (SAM) cards contain VLSI chips implementing APEs and conform to a standard mechanical and interface specification. The number of APEs in a system can be increased by adding more SAMs. The SAM and SAMbus standards allow existing systems to be simply upgraded as new generations of VLSI chips become available.

The SMM-1016 is a stackable SDS memory expansion module that can sustain a 120 Mbytes/s access bandwidth.

2.2 Associative String Processor (ASP)

As mentioned above, the processing core in System-V is an SIMD processing structure implemented using ASP Modules. The ASP, shown in Figure 3, is a modular massively parallel and inherently fault-tolerant processing architecture.

At the logical level, the Associative String Processor (ASP) constitutes a high-performance cellular string associative processor, whereas, at the physical level, the ASP is implemented as a bit-parallel word-parallel associative parallel processor. The ASP is a programmable, homogeneous and fault-tolerant fine-grain SIMD massively parallel processor incorporating a string of identical Associative Processing Elements (APEs), a reconfigurable inter-processor communication network and a Vector Data Buffer for fully-overlapped data input-output as indicated in Figure 3.

Each APE, depicted in Figure 4, incorporates a 64-bit Data Register and a 6-bit Activity Register, a 70-bit parallel Comparator, a single-bit full-adder, 4 status flags and...
control logic for local processing and communication with other APEs. The 6-bit Activity Register is used to select subsets of APEs for subsequent parallel processing.

The APEs are connected via the Inter APE Communication Network, which supports the navigation of data structures and implements a simply-scalable, fault-tolerant and dynamically-reconfigurable configuration to connect APE sources and corresponding APE destinations of high-speed activation signals, implementing a fully-connected dynamically-configured (programmer-transparently) permutation, and broadcast network for APE selection and inter-APE routing functions; synchronous bidirectional multi-bit communication, via a high-speed bit-serial shift register for data/message transfer between APE groups. Thus, the interconnection strategy adopted for the ASP supports a high degree of parallelism for local communication and progressively lower degrees of parallelism for longer distance communication. In particular, the chordal ring topology enables the Inter APE Communication Network to be implemented as a hierarchy of APE groups. Thus, communication times are significantly reduced through automatic bypassing of those groups that do not include destination APEs. In a similar way, namely through bypassing of faulty groups of APEs, fault tolerance of the ASP architecture is guaranteed.

In operation, data are distributed over the APEs and stored in the local Data Registers. Successive computational tasks are performed on the stored data and the results are dumped via the PDX, to the Vector Data Buffer (shown in Figure 3). The ASP supports a form of set processing, in which a subset of active APEs (i.e., those which match broadcast scalar data and activity values) support scalar-vector (i.e., between a scalar and Data Registers) and vector-vector (i.e., within Data Registers) operations. Matching APEs are either directly activated or source inter APE communications to indirectly activate other APEs via the Inter APE Communication Network. The Match Reply (MR) line to the control interface provides feedback on whether none or some APEs match. The APE can operate in three different data modes dictated by the Data Register configuration. The supported modes are: storage and bit-parallel processing of two 32-bit words or four 8-bit byte fields and storage and bit-serial processing of one to three bit-fields of varying length (of no more than 64 bits per field). The instruction set is based on 4 basic operations, match, add, read and write. In order to achieve bit-level masking, during match and write operations, the corresponding byte and bit fields of the Data Bus are represented with ternary (i.e., 2-bit) digits.

3 The random number generator

The random number generator we used here is a lagged Fibonacci, algorithmically the same that was described in [19] in more details. We exploit the orthogonal manipula-
tion capabilities of the ASP string and generate 160 bit long pseudo random numbers along the string with fast "look ahead carry" operations.

\[ x_i = x_{i-17} \pm x_{i-5} \pm c \] (1)

We segment the ASP string to 160 APE parts and so we generate No. of APE-s / 160 such numbers in parallel. We take one bit of the generated numbers for each processor as a bit of a 18-bit integer random number. hence to build up a full 18-bit random number for an APE we have to repeat the \( x_i \) generation step 18 times. The lag columns are moving in a circular way in the 18-bit memory field of APE-s. The carries of eq. (1) are transmitted back between two addition steps to the beginning of the segments by the effective "Activity Link" operation. So we can generate a 18-bit random number – which is a compromise of the on-processor memory and necessary resolution – within a few clock cycles in all APE-s independently of the system size.

These integer random numbers can be thresholded in parallel by constants \( (p \times 2^{18}) \) that are stored in the memory of the ISM. By this operation we can tag processors with probability \( p \). Practically we could achieve a \( \sim 10^4 \) update/sec. that means \( \sim 10^8 \) random numbers in every second if the system size is 12K.

Testing of the random number generator was through comparing the simulation results with results obtained on a serial computer.

4 Time dependent Monte Carlo simulation

Time dependent Monte Carlo simulation suggested by [5] has become a very precise and effective tool in statistical physics. We start the system from a state that is usually random or near to the absorbing state and follow the evolution of its statistical properties for a few thousand time steps.

In general one usually "measure" the following quantities

- survival probability \( p_s(t) \) of the initial seed
- order parameter density \( \rho(t) \)

The evolution runs are averaged over \( N_s \) independent runs for each different value of \( p \) in the vicinity of \( p_c \). At the critical point we expect these quantities to behave in accordance with the power law as \( t \to \infty \), i.e.

\[ p_s(t) \propto t^{-\delta}, \] (2)

\[ \rho(t) \propto t^{\alpha}, \] (3)

For estimating critical exponents and the critical point there is a very effect way by analysing the local slopes of the \( \log - \log \) quantities. Example for \( \rho \)

\[ \alpha(t) = \frac{\ln [\rho(t)/\rho(t/m)]}{\ln(m)} \] (4)

where we use \( m = 8 \) usually. In the case of power-law behaviour we expect \( \alpha(t) \) to be straight line as \( 1/t \to 0 \), when \( p = p_c \). The off-critical curves should possess curvature. Curves corresponding to \( p > p_c \) should veer upward, curves with \( p < p_c \) should veer down.

5 The NEKIM model

The research of phase transitions of non-equilibrium models is in the forefront of statistical physics. Very few models are solved and the universality class picture of equilibrium systems can not be directly transferred to here. The lack of detailed balance condition

\[ P(\{s\})W(\{s\} \to \{s'\}) = P(\{s'\})W(\{s'\} \to \{s\}) \] (5)

— where \( P(\{s\}) \) is the probability of a state, and \( W \) is the transition probability — enables arbitrary noise and this seems to have an effect on the ordered state influencing the critical scaling behaviour. This suggests much richer behaviour than in equilibrium statistical systems. Contrary to this for a long time there has been only one phase transition universality class known according to which models are categorised in equilibrium systems. Namely every continuous phase transition to an absorbing state have ultimately been found to belong to the class of Directed Percolation or Reggeon Filed theory [1]. Theoretical investigations have shown the robustness of this class [6, 7, 8]. There are a few exceptions found up to now. One such an exceptional class is the Parity Conserving (PC) universality class in 1d, which was named after that the number of particles is conserved modulo 2. Later it was realized [3] that more precisely the special dynamics the "Branching and annihilating random walk with even number of offsprings" (BARWe) is responsible for this non-DP behaviour since the underlying field theory possesses a special symmetry in this case. The field theoretical description of this class has not given quantitatively precise results and we can rely on simulation results for critical exponents and scaling relations. We have been investigating one representative of the PC class for some years namely a special non-equilibrium kinetic Ising model in 1d (NEKIM) [8].

The Ising model is the simplest system that is capable of describing of ferro-magnetism on the basis of collective action of spin variables \( (s_i) \). The generalisation of the static model that involves spin flip dynamics was done by Glauber [4] first. The Glauber model is exactly solvable in 1d and the kink variables \( (k_i) \) (‘01’ or ‘10’ pairs ) between the ordered domains has been found to exhibit annihilating
random walk. Other kind of dynamics can also be introduced that lead to the same equilibrium state, for example the spin number conserving Kawasaki dynamics (see [13]). It was suggested that if we apply different kind of dynamics alternately we can create a system that does not have an equilibrium state described by Boltzmann distribution but may possess a steady state in which the some global parameters are constant (magnetisation example) similarly to the eq. systems but others are not (example particle or energy currents can flow through). The alternating application of Glauber spin-flip and Kawasaki spin-exchange was proposed by [13] and it was discovered by [14] that there is non-temperature driven phase transition in which the kinks exhibit BARWe dynamics and so the universality class is PC. Originally this transition has been shown for zero temperature spin-flip (because in 1d any finite temperature causes such a strong fluctuations that destroy the order) plus an infinite temperature (process that does not depend on the local neighbourhood of spins) spin-exchange. The spin exchanges don’t destroy the order because they don’t do anything inside the ordered domains.

In this work we investigate numerically the generalisation of this model for finite temperature spin-exchange ($T_K$) and investigate its effect on the transition [15]. We define the model on the 1d ring (i.e. periodic boundary conditions) with the transition probabilities:

- kink random walk : $w_W = \Gamma (1 - \delta)$
- kink annihilation : $w_A = \frac{\Gamma}{2} (1 + \delta)$
- spin-exchange : $w_{i,i+1} = \frac{\Gamma}{2} (1 - s_i s_{i+1})(1 - \frac{\Gamma}{2}(s_{i-1}s_i + s_{i+1}s_{i+2}))$

where the parameters we used are : $\Gamma = 0.35$, $p_T = \exp(-4J/kT_K)$, $\gamma = (1 - p_T)/(1 + p_T)$, $p_{ex} = 0.239$. The free parameters are : $\delta, T_K$.

We used one site per processor mapping and therefore a parallel updating version of the NEKIM had to be invented in order to exploit the resources of the ASP. To realize the above processes "two lattice update" was employed in case of the spin-flip. That means that in one step the "even" lattice sites and in the next step the "odd" lattice sites were updated. In case of the subsequent spin-exchange a "three lattice update was performed. Further algorithmic details will be presented in the next section.

The time dependent simulations have been performed on a $L = 3040$ size (number of APE-s) ring by starting from random initial states and following the kink density up to $t_{MAX} = 8000$. The number of runs over which the statistical averaging was performed was $3 - 5 \times 10^4$ for each $p_T$ parameter. First the critical point was located for a fixed value of $\delta = -0.362$, $p_{ex} = 0.239$ by varying $p_T$. As one can see on the local slope Figure 5 it is about $p_T = 0.27$, because the other curves corresponding to other $p_T$-s deviate from scaling as $t \to \infty$. We can read off from the ordinate of the graph that the corresponding critical exponent is $0.28(1)$ that is in a good agreement with the PC class value [10].

We performed timing measurements on a System-V with 12K APEs and with 4 Mbytes of memory (that occupy 3 6U VME slots) and is attached to a SparcStation 5/64 that serves as host for the applications. On that machine $\sim 2 \times 10^{-8}$ second is necessary to update a site. In comparison the serial simulation Fortran program that was run on a DEC-ALPHA 2000 station has achieved $2 \times 10^{-6}$ sec. / site speed.. The other thing that we investigated is the effect of long-range initial conditions that has just recently been shown to be relevant [7] in case of DP transitions. The initial states with $< k_i k_{i+x} > \propto x^{-(1-\sigma)}$ kink-kink correlations and even numbered kinks are generated by the same serial algorithm as described and numerically tested in ref. [7] in the $\sigma \in (0, 1)$ interval. We required the even-numbered initial kink sector because the kink number is conserved mod 2. The spin states are assigned to the kink configuration and are loaded to the ASP string time to time during the trial runs. The kink density has been measured in $L = 12000$ sized systems up to $t_{max} = 80000$ time steps such that we can observe good quality of scaling for three decades in the $(80, 80000)$ time interval (see Figure 6). As one can see there is an increase with exponent $\alpha \sim 0.28$ in the kink density for $\sigma = 0$, where in principle only one

![Figure 5. Local slopes of $\log(\rho_{kink}(t))$ versus $\log(1/t)$ in the NEKIM simulations at for $p_T = 0.26, 0.265, 0.27, 0.275, 0.28, 0.285, 0.29$ (from bottom to top curves). Scaling can be observed for $p_T = 0.27(1)$, with the exponent $\alpha \sim 0.28$ in agreement with PC order parameter density decays.](image)
pair of kinks is placed on the lattice in agreement with former simulation results [18]. On the other extreme case for $\sigma = 1$ the kink density decays with $\alpha \sim -0.28$ exponent again in agreement with our expectations i [16]. In between the exponent $\alpha$ changes continuously as the function of $\sigma$ and changes sign at $\sigma = 0.5$. That means that the state generated with $\sigma = 1/2$ is very near to the $t \to \infty$ steady state limit [20].

![Figure 6. log($\rho_{\text{kink}}(t)$) versus log($t$) in the NEKIM simulations at for $\sigma = 0, 0.1, 0.2..., 1$ initial conditions (from bottom to top curves).](image)

6 Monte Carlo simulation algorithm

The low-level ASP part of the simulation is similar to what was described in [19]. We map the 1d system onto a non-segmented string (left and right ends of the string are connected) (but in case of random generation steps we re-segment it to 160 PE substrings to avoid long communications). The left and right neighbour informations are shifted simultaneously to each APE and we apply the transition rules with ternary masks and the random threshold conditions. Since we update in two (and three) sub-lattices every second and third APE are marked by Activity bits (A,B,C) that is fixed from the beginning of the simulations and we take into account these Activity bit conditions as well when doing spin-flip (by the ‘Add’ operation) or the spin exchange. See Figure 7 for the APE representation of the kink annihilation process. For kink random walk the only difference is that we use different masks (100 or 110 ... etc.). In case of the finite temperature spin-exchange we have to take into account even more conditions (i.e. four spin states) as shown on Fig. 8. The spin state is transformed to kink state within a single (Shift+Add) operation cycle and an internal “Global Match reply” mechanism examines if the system evolves to the kink-free ordered state. If it is in the ordered state the whole time development cycle stops, because in the absorbing state there are no more changes take place. The entire kink state is dumped by the efficient PDX exchange within a few clock cycles to the PDS where a built in hardware mechanism counts the number of ‘1’-s overlapped in time with the MC updates. We invoke this dumping in every 8 MC updates only to keep a time balance with the counting.

The site update time in this case slightly smaller than in case of the finite temperature NEKIM : $10^{-8}$ second. The serial version of the program run on a SUN station with UltraSparc 300 MHz processors achieved $8 \times 10^{-7}$ site update speed. That means that the System-V algorithm is 80 faster. We made a test run on the FUJITSU AP3000 parallel supercomputer with $n = 16$ nodes as well where the parallelism has been exploited on the level of trial samples. We found that the System-V program is a factor of 5 faster than the simulation on the AP3000.

7 Conclusions

In this work we have shown that the System-V image processing architecture is capable for Monte Carlo simulation with a performance far most exceeding present supercomputing technologies if an effective on-processor random generator is invented. This high performance is possible
addressable, associative string processing. That means that owing to the data movement minimisation of the content parallel data exchanger PDX does not interrupts the internal (like measurements on the state at time to time) the fast and in the next generations) APEs.

Therefore we think that this kind of MPC SIMD modules are good candidates for building blocks in large scale, “Grand Challenge” simulation architectures.

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References

[1] J. L. Cardy and R. L. Sugar. J. Phys. A: Math. Gen., 13:L423, 1980.
[2] J. L. Cardy and U. Tauber. Phys. Rev. Lett., 77:4780, 1996.
[3] M. Droz, Z. Rácz, and J. Schmidt. Phys. Rev. A 39, 39:2141, 1989.
[4] R. J. Glauber. Phys. Rev. Lett., 68:3060, 1992.
[5] P. Grassberger. Z. Phys. B, 47:365, 1982.
[6] P. Grassberger and A. de la Torre. Ann. of Phys, 122:373, 1979.
[7] H. Hinrichsen and G. Ödor. Correlated initial conditions in directed percolation. Phys. Rev. E., 58:311, 1998.
[8] H. Hinrichsen and G. Ödor. Critical behavior of roughening transitions in parity-conserving growth processes. Phys. Rev. E, 60(4), October 1999.
[9] H. Hinrichsen and G. Ödor. Roughening transition in a model for dimer adsorption and desorption. Phys. Rev. Lett., 82:1205, 1999.
[10] H. K. Janssen. Z. Phys. B, 42:151, 1981.
[11] D. Kastenbaum. A perfect match. New Scientist, April 1998.
[12] K. Kawasaki. Phase Transition and Critical Phenomena, 2:443, 1972.
[13] A. Krikelis and R. M. Lea. A modular massively parallel computing approach to image-related processing. proceedings of IEEE , 84(7), July 1996.
[14] R. M. Lea. Asp: a cost effective parallel microcomputer. IEEE Micro, October 1988.
[15] N. Menyhárd. privat communication.
[16] N. Menyhárd. J. Phys. A : Math. Gen., 27:6139, 1994.
[17] N. Menyhárd and G. Ödor. Non-equilibrium phase transitions in one-dimensional kinetic ising models. J. Phys. A : Math. Gen., 28:4505, 1995.
[18] N. Menyhárd and G. Ödor. Phase transitions and critical behaviour in one-dimensional non-equilibrium kinetic ising models with branching annihilating random walk of kinks. J. Phys. A : Math. Gen., 29:7739, 1996.
[19] G. Ödor, G.Vesztergombi, and F.Rohrbach. Parallel simulation of 1d probabilistic cellular automata. Proc. of the Fifth Euromicro Workshop on Parallel and Distributed Processing, pages 149–154, 1997.
[20] G. Ödor and N. Menyhárd. in preparation.
[21] F. Rohrbach. The mpcc project : final report. CERN preprint, 93(7), 1993.