Simulation on High-Performance Computers

Stefan Lang

Interdisciplinary Center for Scientific Computing (IWR) University of Heidelberg INF 368, Room 532 D-69120 Heidelberg phone: 06221/54-8264 email: Stefan.Lang@iwr.uni-heidelberg.de

WS 15/16

Organisational Information

- Course title: Parallel High Performance Computing
- Lecturer: Stefan Lang, Scientific Computing, IWR, Room 425
- Course extend: 4h lectures + 2h exercises, CP: 8
- In total 30 lectures and 12 exercise sheets
- Dates
 - Lectures: Tu 9.00-11.00 (V R532), Th 9.00-11.00 (V R532)
 - Exercises: Th 14.00-16.00 (E in CIP Pool, OMZ, INF 350 U.012)
- Prerequisites:

Basic lectures in Computer Science and Numerics

- Helpful: Knowledge of C/C++
- Literature: see course homepage http://conan.iwr.uni-heidelberg.de/teaching/phlr_ws2015/index.html

What means Scientific Computing?

Scientific Computing is a rapdily growing novel discipline of science

In particular Numerical Simulation (NS):

- Goal of NS is to simulate natural or technical processes with computing machines
- Approach accross disciplines: natural scientists, engineers, mathematicians and computer scientists, all need to work together
- Problems, that are relevant in practice, are handled systematically with formal methods
- NS enables insight into areas that are difficult to access in lab experiments and field studies, for example neuroscience, cell biology, water economics, astrophysics

Why Scientific High-Performance Computing?

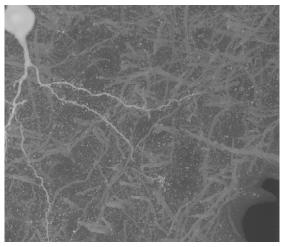
Trends in Numerical simulation:

- Treatment of global models instead of local partial models (Models of entire brain areas, Complete water resource systems, virtual prototyping in aircraft, naval and automobile design)
- Analysis of coupled overall systems instead of isolated individual processes (multi-media, multi-phase, multi-scale processes, convection-diffusion-reaction)

• Computer gets a **scientific observation instrument** (high resolution capacity, measurement in some areas difficult/impossible, parameter studies performable)

Computational Neuroscience - Signal Processing

 Goal: Development of a neuronal network model, that reflects an observed or measured system behaviour



Simulation of neuronal network models, statistical analysis of realisations

A Supercomputer

ASCI Red Storm

Successor of the first TeraFlop computer ASCI Red 1997 Hardware:

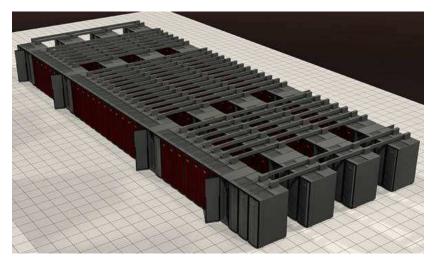
- 11.646 × 2 GHz AMD Opteron CPUs
- CPU boards vertically mounted in 108 cabinets
- 4 GFLOPS per CPU (40 TFLOPS total)
- I GB per CPU (10 TB total)
- Shared memory inside the node
- 3D mesh full interconnect

Software:

- Compute nodes: custom Sandia-developed light-weight OS code-named Catamount
- Service and storage nodes SuSE Linux.

A Supercomputer

ASCI Red Storm



Scalability

Algorithmic complexity using the example of linear solvers

for an equation of the form

$$Ax = b$$

Dimension	d = 2	d = 3
Gaussian elimination	$O(N^3)$	$O(N^3)$
Banded Gauss	$O(N^2)$	$O(N^{2.33})$
Nested Disection Gauss	$O(N^{1.5})$	$O(N^2)$
Richardson, GS, Jacobi	$O(N^2)$	$O(N^{1.67})$
CG, SOR	$O(N^{1.5})$	$O(N^{1.33})$
SSOR–CG	$O(N^{1.25})$	$O(N^{1.17})$
Multigrid	O(N)	O(N)
Cascadic Iteration	O(N)	O(N)

Scalability

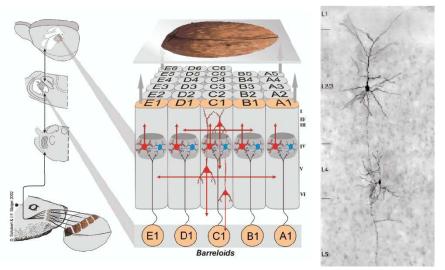
Scalability means in simple words A problem of steadily increasing size can be calculated on a steadily increasing computer (nearly) equally fast.

This is seldom the case (unfortunately)!

As a consequence we need

- enabling software to handle such powerful computers
- scalable algorithms + scalable implementations + scalable architectures
- ... buying huge and expensive computers alone is not enough!!

A Biological System: The Barrel Cortex of the Rat



Goal: Mechanistic understanding of easy decision making

Modeling of Neuronal Activity

Hodgkin-Huxley equation:

The electric potential $v(\mathbf{x}, t)$ and gating particles $\mathbf{c}(\mathbf{x}, t) = (m(\mathbf{x}, t), h(\mathbf{x}, t), n(\mathbf{x}, t))^T$ obey

$$\begin{split} \mathbf{c}_{m}\partial_{t}\mathbf{v} &= \partial_{\mathbf{x}}\mathbf{g}_{a}\partial_{\mathbf{x}}\mathbf{v} + i_{nj} - \sum_{\nu \in \mathcal{C}} i_{\nu}(\mathbf{v}, \mathbf{c}) - i_{s}(\mathbf{v}) \\ \partial_{t}\mathbf{c}_{\mu} &= \alpha_{\mu}(\mathbf{v}) \cdot (\mathbf{1} - \mathbf{c}_{\mu}) - \beta_{\mu}(\mathbf{v}) \cdot \mathbf{c}_{\mu}, \end{split}$$

with $\nu \in \mathcal{C} =: \{Na, L, K\}$ and $\mu \in \{m, h, n\}$. Boundary and initial conditions are given by

$$\begin{split} g_a \partial_{\mathbf{x}} v &= g_N & \text{ on } \partial \Omega_N, \\ v &= g_D & \text{ on } \partial \Omega_D, \\ (v, \mathbf{c})(\mathbf{x}, 0) &= (v^0, \mathbf{c}^0) \text{ for } t = 0. \end{split}$$

The ion currents are modeled with the gating particles and are given by

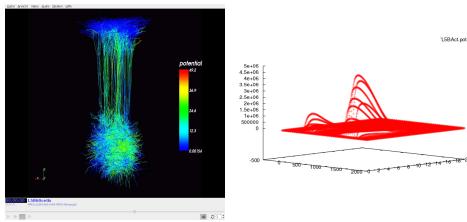
$$\begin{split} i_{Na}(v) &= \overline{g_{Na}} \cdot m^3 h \cdot (v - E_{Na}), \\ i_K(v) &= \overline{g_K} \cdot n^4 \cdot (v - E_K), \\ i_L(v) &= \overline{g_L} \cdot (v - E_L) \end{split}$$

Moreover currents of synapses are modeled by

$$i_{\rm s}(v,t)=g_{\rm s}(t)\cdot(v-E_{\rm s})$$

with time-dependent synaptic strength g_s .

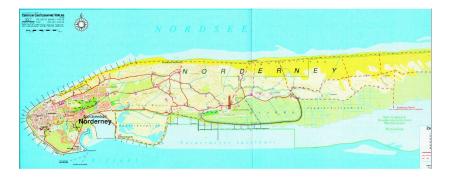
Simulation Study: Passive Deflection of a Whisker

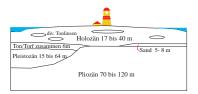


60 L5B neurons activated by VPM

Spatial and temporal activity distribution

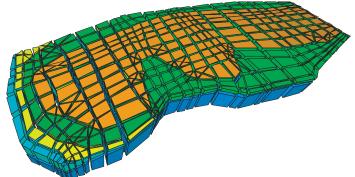
Problem study Norderney





- Project "coastal preservation"
- Island typical fresh water lense (-85m)
- Simulation of lense constitution and water facilitation from two pumps
- Strategy to preserve the water quality

Problem study Norderney



10km x 4km x 150m

- Initial grid (1516 elements, D. Feuchter, geomod2ng)
- 6 geological layers with varying permeability 10⁻¹⁰ 10⁻¹⁵
- Boundary conditions: influence of fresh water on upper boundary wells are sinks, in/outflow in coastal areas, hydrostatic pressure

Density-driven Groundwater Flow

The equations of density-driven flow are derived from conservation laws. Formulation uses salt mass fraction ω and pressure p

$$\partial_t(n\rho) + \nabla \cdot (\rho \mathbf{v}) = \mathbf{Q}\rho, \qquad (f \text{ low})$$
$$\partial_t(n\rho\omega) + \nabla \cdot (\rho \mathbf{v}\omega - \rho D\nabla\omega) = \mathbf{Q}\rho\omega \quad (transp.)$$

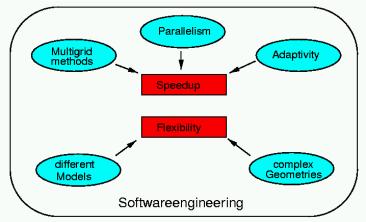
with

$$\begin{split} \mathbf{v} &= -\mathbf{K}/\mu(\nabla \mathbf{p} - \rho \mathbf{g}), \qquad \text{(Darcy's law)} \\ D &= (\alpha_L - \alpha_T)\mathbf{v}/|\mathbf{v}| + \alpha_T |\mathbf{v}| \qquad \text{(Scheidegger)} \end{split}$$

Proper initial and boundary conditions have to be defined.

Method and Software Development

Moreover realistic problems are difficult to handle



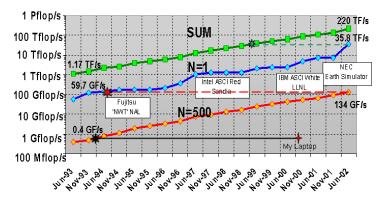
The realization of all these aspects together is a difficult task!

Software engineering is not subject of this course.

TOP500 - List of SuperComputers

TOPSOO

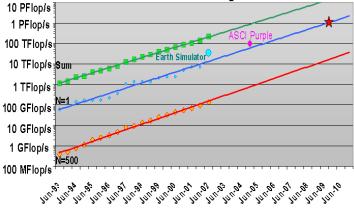
TOP500 - Performance



TOP500 - Trend

TOPSOO

Performance Extrapolation



Petaflop machine in 2010! Exaflop until 2018?

Parallelisation: An Introductory Example

Scalarproduct of two vectors

- Introduction of an adequate notation
- Interaction via shared variables
- Interaction via passing of messages
- Evaluation of parallel algorithms

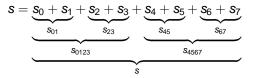
A Simple Problem: Scalar Product Generation

Scalar product of two vectors of length *N*:

$$s = x \cdot y = \sum_{i=0}^{N-1} x_i y_i.$$

Parallelisation idea:

- Summands x_i and y_i are independent
- ② $N \ge P$, form $I_p \subseteq \{0, ..., N-1\}$, $I_p \cap I_q = \emptyset \forall p \ne q$ Each processor calculates now the partial sum $s_p = \sum_{i \in I_p} x_i y_i$
- Summation of partial sums in example for P = 8:



Fundamental Conceptions

- Sequential program: Sequence of instructions that are processed sequentially.
- Sequential process: Active execution of a sequential program.
- Parallel computation: Set of interacting sequential processes.
- Parallel program: Describes parallel calculation. Given by a set of sequential programs.

Notation for Parallel Programs I

- Preferably simple and detached of practical details
- Allows different programming models

```
Program (Pattern of a parallel program)
parallel < program name>
    // section with global variables (accessible by all processes)
    process < processname-1> [< copyparameters>]
        //local variables, that can be read and written
        // by process < Prozessname-1 > only
        // Applications in C-like syntax. Mathematical
        // formula or text allowed for simplification.
    process < processname-n> [< copyparameters>]
```

Notation for Parallel Programs II

Variable declaration

double x, y[P];

Initialisation

int $n[P] = \{1[P]\};$

Local/global variables

The execution of the parallel computation starts with initialisation of global variables, then the the processes are executed

Remarks regarding process term: shared variables

Scalarproduct with Two Processes

Program (Scalarproduct with two processes)

```
parallel two-process-scalar-product
     const int N=8:
                                             // problem size
     double x[N], y[N], s=0;
                                             //vectors.result
     process Π1
           int i:
           double ss=0:
           for (i = 0; i < N/2; i++)
                ss += x[i]^*v[i]:
                                             // danger!
           s=s+ss;
     process П<sub>2</sub>
           int i:
           double ss=0;
           for (i = N/2; i < N; i++)
                ss += x[i]^*v[i]:
                                             // danger!
           s=s+ss;
     }
```

Variables are global, each process works on a part of the indices

Collision during write access!

Critical Section I

 High level language instruction s = s + ss is transformed in assembly instructions:

Process Π₁

- 1 load s into R1 load ss into R2 add R1 and R2 result in R3
- 2 store R3 into s

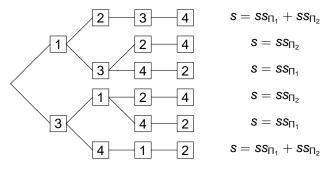
Process Π₂

- 3 load s into R1 load ss into R2 add R1 and R2 result in R3
- 4 store R3 into s
- Execution sequence of instructions of different processes is not determined.

Critical Section II

Possible execution sequences are:

Result of calculation



• Only sequences 1-2-3-4 and 3-4-1-2 are correct.

Critical Section III

- Instruction block builds a critical section, that needs to be processed by mutual exclusion.
- We quote this *at first* with squared brackets

```
[ (instruction 1); ...; (instruction n); ]
```

- The symbol "[" selects a process to work on the critical section, all others are waiting.
- Efficient realisation requires hardware instructions, that are introduced later.

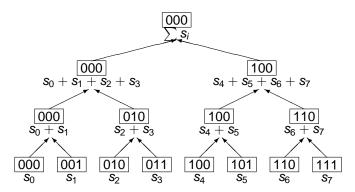
Parameterisation of Processes

- Processes contain partly identical code (using different data)
- Parameterise the code with a process number, choose the data to be processed using this number
- SPMD = single program multiple data

```
Program (Scalarproduct with P processors)
parallel many-process-scalar-product
ł
   const int N:
                               // problem size
   const int P:
                            // process count
   double x[N], y[N]; // vectors
                  // result
   double s = 0:
   process \Pi [int p \in \{0, ..., P - 1\}]
        int i: double ss = 0:
        for (i = N * p/P; i < N * (p + 1)/P; i++)
           ss += x[i] * y[i];
        [s = s + ss]:
                     // Here still all are waiting again
```

Communication in Hierarchical Structure

Treelike organisation of the communication sequence with Id P levels



In level i = 0, 1, ...

- Processes, whose last i + 1 bits are 0, fetch
- results of processors whose last *i* bits are 0 und whose bit *i* is 1

parallel parallel-sum-scalar-product

const int d = 4; const int N = 100; const int $P = 2^d$; double x[N], y[N]; double $s[P] = \{0[P]\}$; int $flag[P] = \{0[P]\}$;

// problem size // process count // vectors // result // process p is ready

process Π [int $p \in \{0, ..., P-1\}$] { int *i*, *r*, *m*, *k*;

> for (i = N * p/P; i < N * (p + 1)/P; i++)s[p] + = x[i] * y[i];

for
$$(i = 0; i < d; i++)$$

{
 $r = p \& \left[\sim \left(\sum_{k=0}^{i} 2^{k} \right) \right];$
 $m = r \mid 2^{l};$
if $(p == m)$ flag[m]=1;
if $(p == r)$
{
while (!flag[m]);
 $s[p] = s[p] + s[m];$

// delete last i + 1 bits

// set bit i

// conditional synchronisation

}

Parallelisation of Summation II

- New global variables: *s*[*P*] partial results *flag*[*P*] indicates, that a processor has finished its computation
- Waiting is called *conditional synchronisation*
- In this example mutual exclusion could be exchanged by conditional synchronisation. This does not work always!
- Reason is that we have fixed the order in advance

Localisation

Goal: Avoid global variables We advance in two steps: (I) Localise vectors x, y, (II) localise result s

```
Program (Scalarproduct with local data)
parallel local-data-scalar-product
     const int P, N;
     double s = 0:
     process \Pi [ int p \in \{0, ..., P-1\}]
          double x[N/P], y[N/P]; // Assumption N is divisible by P
                                                // Local section of vectors
          int i:
          double ss=0:
          for (i = 0; i < (p + 1) * N/P - p * N/P; i++) ss = ss + x[i] * v[i]:
          [s = s + ss;]
     }
```

Each stores only N/P indices (one more if not exactly divisible), these start always with local number 0

Each local index x equates to a global index in the sequential program:

$$i_{
m global}(p) = i_{
m lokal} + p * N/P$$

Message Passing I

To completely avoid global variables we need a new concept: messages

```
Syntax:

send(<Process>,<Variable>)

receive(<Process>,<Variable>)
```

Semantics:

send operation sends content of variable to the specified process, **receive** operation waits for message of the specified process and copies it to the variable

send operation waits until the message is received successfully, **receive** operation blocks the process until the message is received

Blocking, or synchronous communication (later other)

Message Passing II

Program (Scalarproduct with message passing)

```
parallel message-passing-scalar-product
      const int d, P=2^d, N;
                                                         // constants!
      process \Pi [int p \in \{0, ..., P-1\}]
            double x[N/P], y[N/P];
                                                                      // local section of vectors
            int i.r.m:
            double s, ss = 0:
            for (i = 0; i < (p + 1) * N/P - p * N/P; i++) s = s + x[i] * y[i];
            for (i = 0; i < d; i++)
                                                                      // d steps
                  r = p \& \left[ \sim \left( \sum_{k=0}^{i} 2^k \right) \right];
                   m = r \mid 2^{\overline{i}}
                  if (p == m)
                         send(\Pi_r,s);
                   if (p == r)
                         receive(\Pi_m,ss);
                         s = s + ss:
                   }
            }
```

Evaluation of Parallel Algorithms I

Here: asymptotic behaviour in dependance of problem size and processor count

Sequential runtime:

$$T_{s}(N) = 2Nt_{a},$$

ta: Time for arithmetic operations

Parallel runtime of message-passing variant:

$$T_{\mathcal{P}}(N, P) = \underbrace{2 \frac{N}{P} t_a}_{\text{local scalarproduct}} + \underbrace{\operatorname{Id} P(t_m + t_a)}_{\text{parallel sum}},$$

t_m: time to send a number

speedup:

$$S(N, P) = \frac{T_s(N)}{T_p(N, P)} = \frac{2Nt_a}{2\frac{N}{P}t_a + Id P(t_m + t_a)}$$
$$= \frac{P}{1 + \frac{P}{N} Id P \frac{t_m + t_a}{2t_a}}$$

It holds $S(N, P) \leq P$!

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Evaluation of Parallel Algorithms II

Efficiency:

$$E(N,P) = \frac{S(N,P)}{P} = \frac{1}{1 + \frac{P}{N} \operatorname{Id} P \frac{t_m + t_a}{2t_a}}$$

It applies $E \leq 1$

Consideration of asymptotic limit cases:

- Fixed *N*, growing *P*: $\lim_{P\to\infty} E(N, P) = 0$
- Fixed *P*, growing *N*: $\lim_{N\to\infty} E(N, P) = 1$ For which relation of $\frac{P}{N}$ "acceptable" efficiency values are achieved is regulated by the factor $\frac{t_m+t_a}{t_a}$. This is the relation of communication to computation time.
- Scalability for simultaneously growing of *N* and *P* in the form N = kP:

$$E(kP,P) = \frac{1}{1 + \operatorname{Id} P \frac{t_m + t_a}{2t_a k}}$$

Drops off slowly with $P \rightarrow$ good scalable.

Exemplary for many algorithms!

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Topics

Hardware

- (Parallel) computer architecture: SMP, vector computer, parallel computer
- realisations: some architectures in detail (Paragon, ASCI Red Storm, IBM Blue Gene)
- Programming models
 - Programming models: OpenMP, MPI, PThreads
 - Fundamental numerical algorithms: matrix multiplication

Algorithms

- Performance evaluation: Efficiency, speedup, scalability
- Parallel sorting
- Dense und sparse filled equation systems

Applications

 Parallel (Numerical) applications (neuroscience, biology, soilphysics, astrophysics)

Themes are without commitment.

Organisational Stuff

- Exercises Thursday 14.00-16.00
 Theme: Introduction to (Advanced) C++
 CIP Pool, OMZ, INF 350, U.012
- Certificate: regular participation, 50% points
- Exam: mostly oral after consultation
- Course count: 8 credit points