# Load Balancing for Parallel Multi-core Machines with Non-Uniform Communication Costs

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# Load Balancing for Parallel Multi-core Machines with Non-Uniform Communication Costs

Laércio Lima Pilla, Christiane Pousa Ribeiro, Daniel Cordeiro, Chao Mei, Abhinav Bhatele, Philippe O. A. Navaux, François Broquedis, Jean-François Méhaut, Laxmikant V. Kale, Pierre Coucheney, Bruno Gaujal



# Agenda

#### Introduction

- Applications
- Parallel Machines
- Issues and Objectives

#### Load Balancing Approach

- General Idea
- Required Information
- NucoLB
- TopoAwareLB

#### Performance Evaluation

- Platforms
- Load balancers
- Benchmarks and Application Results

#### Concluding Remarks

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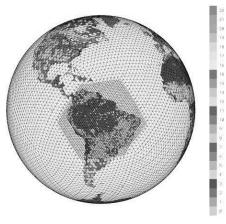
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- Load balancers
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#### Concluding Remarks

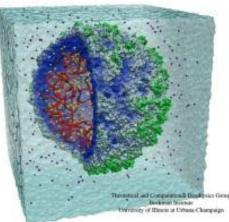
**Applications** 

### Irregular Applications

- Unpredictable, dynamic behaviour
- Input-dependent
- Compute intensive (HPC)
- Iterative simulations
- Load imbalance
- Complex communication patterns



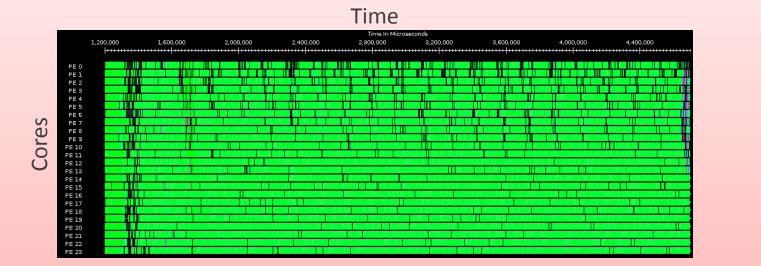
Climatology



#### **Molecular Dynamics**

**Applications** 

#### Irregular Applications



Example: LeanMD on a 48 cores machine (24 cores shown)

- Molecular dynamics application written in Charm++
- 1875 tasks for 3.5 seconds (20 iterations)
- Average core usage: 87.5%
- Worst core: 65% usage

**Applications** 

#### Irregular Applications

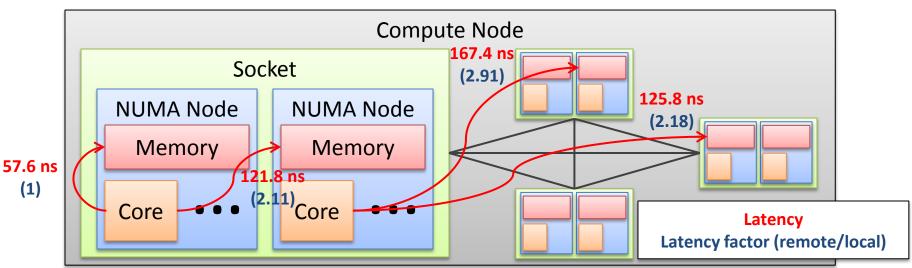


Example: LeanMD on a 48 cores machine (24 cores shown)

- Molecular dynamics application written in Charm++
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#### **Parallel Machines**

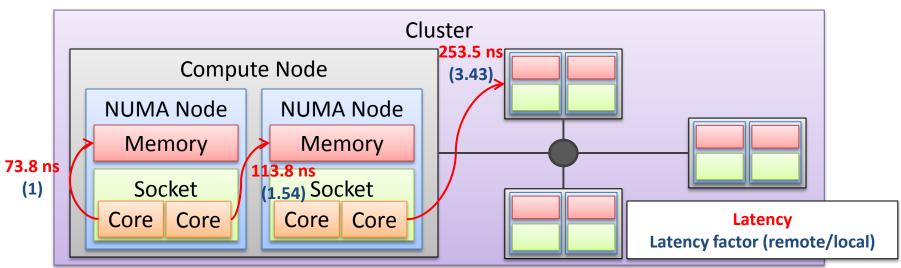
- [Highly] Hierarchical Architectures
  - Memory and network hierarchies
  - Asymmetric communication costs
    - Latency, bandwidth
  - Component sharing



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#### **Parallel Machines**

- [Highly] Hierarchical Architectures
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**Issues and Objectives** 

#### Issues

- Imbalanced processors (idleness)
- Communication overheads (lack of affinity)
- Opposing problems
  - Improving one may worsen the other

**Issues and Objectives** 

#### Objectives

- Improve performance
- Optimize resource usage
  - Reduce processor idleness
  - Reduce communication costs
  - Find the best trade-off
- Performance portability
  - Different platforms, different applications

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**General Idea** 

#### General Ideas

Combine knowledge about

- Application
- Machine topology
- Load balancing
  - Change task mapping dynamically
- Avoid task migrations
  - Data movement costs

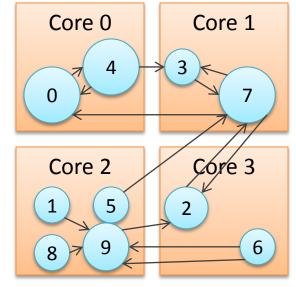
**Required Information** 

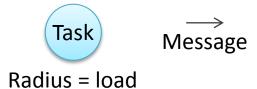
### Discover the Application Behaviour

#### - Profile the parallel execution

- Runtime system level
- Collect

- Task computational loads
  - Execution times
- Communication graph
  - Number of messages between tasks, bytes
- Current task mapping





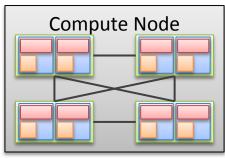
**Required Information** 

7

### Explore Platform Topology

#### Cluster configuration

- Network map, routers, nodes
- Intra-node hierarchy
  - Sockets, cores, NUMA nodes, caches, memory
- Benchmarked communication costs
  - Imbench, stream, MPI ping-pong
  - Latency, bandwidth, derivatives



Latency matrix for NUMA nodes (ns)

0	57.67	121.8	126.9	167.4	126.7	166.3	125.8	165.2
1	122.4	57.22	126.7	166.8	167.6	126.6	166.2	124.3
2	127.0	125.9	57.86	121.8	125.2	124.3	127.1	166.0
3	167.6	165.5	121.6	57.05	126.0	124.9	167.5	125.6
4	126.4	165.5	124.9	125.3	57.46	121.3	126.3	126.1
5	167.6	126.2	124.7	125.1	122.2	57.18	167.2	166.0
6	125.6	165.4	127.0	165.6	126.4	165.9	58.40	121.7
7	165.4	124.3	168.5	126.2	127.2	168.0	122.6	57.17

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Algorithms

- Load Balancing Algorithm
  - Dynamically correct imbalance
  - Improve communications
  - Current work
    - NucoLB

- Load balancer for machines with non-uniform communication costs
- MultiCoreLB
  - Considers cache hierarchy on its decisions
- TopoAwareLB
  - Provides optimality guarantees

NucoLB

#### NucoLB

- Non-uniform communication costs load balancer
- Greedy list scheduling algorithm
  - Map the task with the biggest execution time to the core with the smallest cost
- Integrated on Charm++
  - Charm++ Parallel System UIUC
  - C++-based parallel programming model and RTS
  - Mature load balancing framework
    - Load balancing plugins
    - Measurement-based
  - Tasks = objects = chares

NucoLB

#### NucoLB

#### – Application information:

- Current task mapping
- Task computational load
- Communication graph (number of messages)
- Machine topology model:
  - Hierarchy
  - NUCO factor (latency factor)
    - NUMA factor inside a machine
    - Network factor in clusters
    - Pre-computed in the target platform

#### **NucoLB**

### NucoLB: algorithm

Input: T set of tasks, C set of cores,

#### M mapping of tasks to cores Output: M' mapping of tasks to cores 1. $M' \leftarrow M$

2. while  $T \neq \emptyset$  do

- **3.**  $t \leftarrow k \mid k \in \arg \max_{l \in T} \operatorname{load}(l)$
- $4. \quad T \leftarrow T \setminus \{t\}$
- 5.  $c \leftarrow p, p \in C \land \{t,p\} \in M$
- 6.  $\operatorname{load}(c) \leftarrow \operatorname{load}(c) \operatorname{load}(t)$
- 7.  $M' \leftarrow M' \setminus \{(t, c)\}$

8. 
$$c' \leftarrow p \mid p \in \arg \min_{q \in C} \operatorname{cost}(q, t)$$

- 9.  $\operatorname{load}(c') \leftarrow \operatorname{load}(c') + \operatorname{load}(t)$
- 10.  $M' \leftarrow M' \cup \{(t, c')\}$

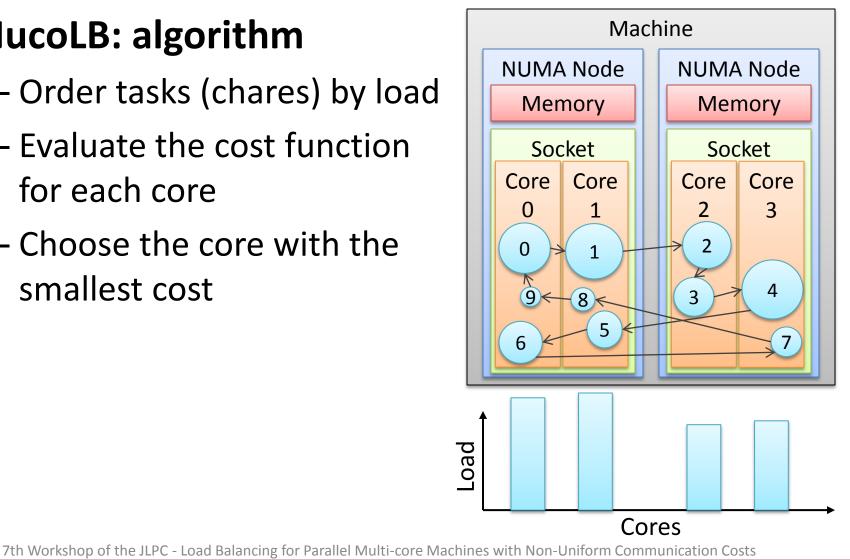
#### where:

- *t*, *k*, and *l* are tasks;
- *c*, *c*', *p*, and *q* are cores;
- load(*t*) is the execution time of *t* in seconds;
- load(c) is the sum of the execution times of the tasks mapped to c in seconds;
- cost (q, t) is the cost to map t to q, as cost(q, t) = load(q) + α\* [r<sub>comm</sub>(t, q) \*NucoFactor(comm(t), node(q)) - l<sub>comm</sub>(t, q)]
- l<sub>comm</sub>(t, q) is the number of local messages sent to t by tasks on the NUMA node of q;
- r<sub>comm</sub>(t, q) is the number of remote messages sent to t by tasks on other NUMA nodes; and
- NucoFactor(comm(*t*), node(*q*)) represents latency factors between the NUMA node of *q* and other NUMA nodes of tasks that communicate with *t*.
- $\boldsymbol{\alpha}$  is a constant to weight the communications

**NucolB** 

### NucoLB: algorithm

- Order tasks (chares) by load
- Evaluate the cost function for each core
- Choose the core with the smallest cost



**NucoLB** 

### • NucoLB: algorithm

- Order tasks (chares) by load

- Evaluate the cost function for each core
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6

3

5

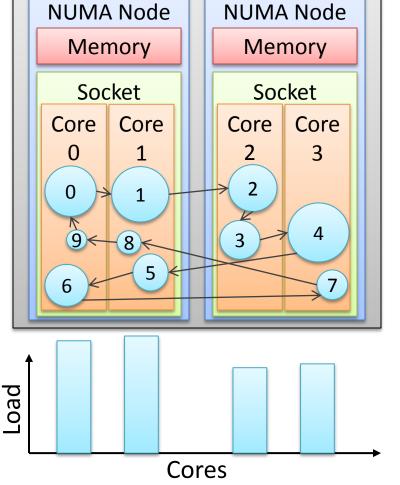
2

0

4

21

1



Machine

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8 9

**NucoLB** 

NUMA Node

### • NucoLB: algorithm

- Order tasks (chares) by load

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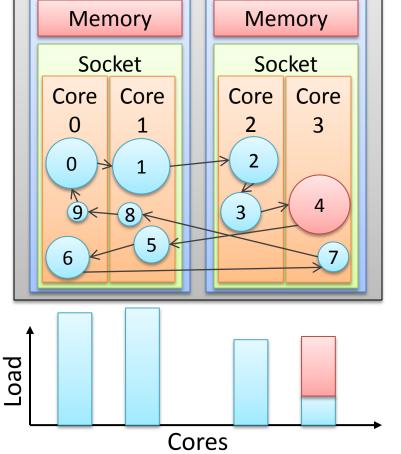
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Machine

NUMA Node

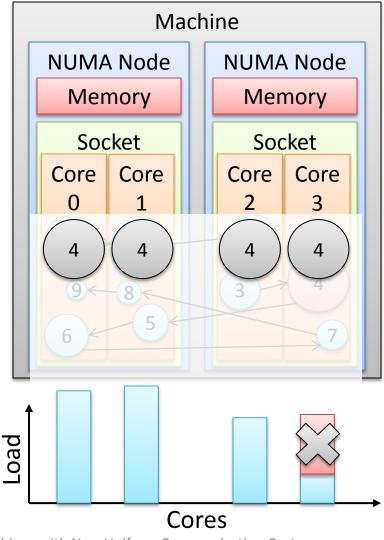
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**NucoLB** 

### NucoLB: algorithm

- Order tasks (chares) by load
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**NucolB** 

NUMA Node

Memory

Socket

Core

3

Δ

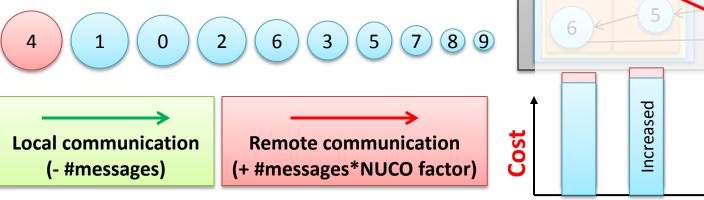
Core

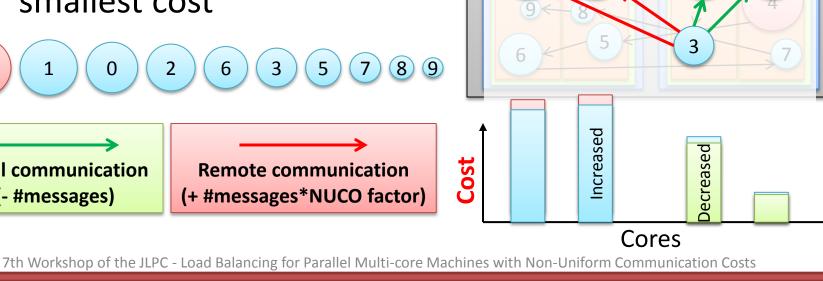
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Machine

### NucoLB: algorithm

- Order tasks (chares) by load
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NUMA Node

Memory

Socket

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Core

**NucoLB** 

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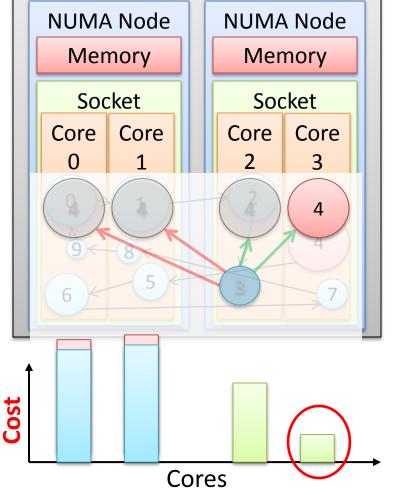
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4



Machine

**NucoLB** 

NUMA Node

### NucoLB: algorithm

- Order tasks (chares) by load
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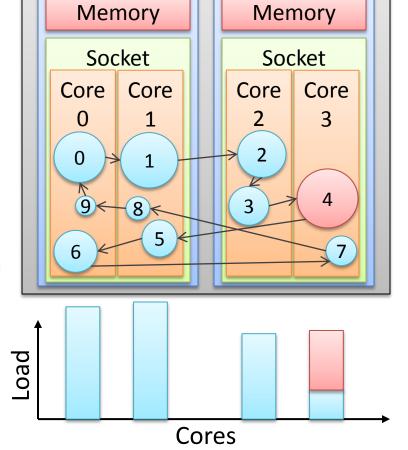
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4

26



Machine

NUMA Node

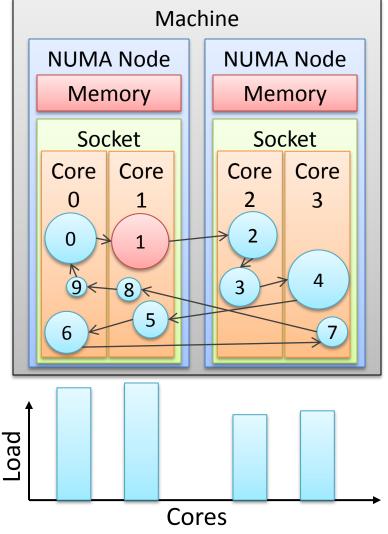
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**NucoLB** 

### NucoLB: algorithm

- Order tasks (chares) by load
- Evaluate the cost function for each core
- Choose the core with the smallest cost



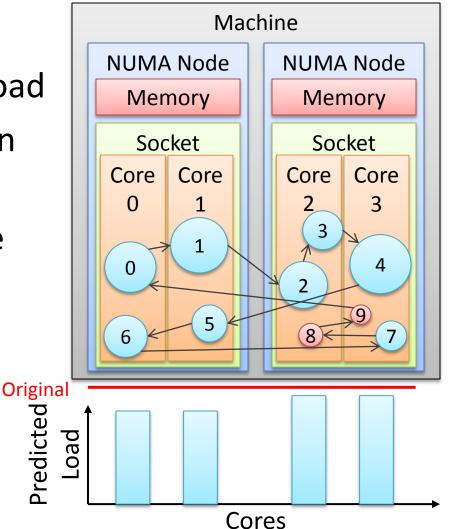
**NucoLB** 

### NucoLB: algorithm

- Order tasks (chares) by load
- Evaluate the cost function for each core
- Choose the core with the smallest cost

The new mapping may change the load of the tasks due to a reduction of communication costs

We usually have more parallelism to work with



**TopoAwareLB** 

#### TopoAwareLB

- Load balancer for multi-core machines
- UMA and NUMA
- Considers cache and memory latencies
- Proved asymptotically optimal
- 24% average improvement over other LBs with benchmarks

**TopoAwareLB** 

#### TopoAwareLB: algorithm

- Choose most loaded core with probability  $\alpha$ 
  - Choose a random core otherwise
- Choose heaviest task in the core with probability  $\boldsymbol{\beta}$ 
  - Choose a random task otherwise
- Choose a new mapping according to a Gibbs distribution with temperature T

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**Parallel Platforms** 

### Platforms

- NUMA32 (UJF-CEA)
  - 4 Intel Xeon processors, 8 cores each
  - 1 NUMA node per socket
- NUMA48 (UJF-CEA)
  - 4 AMD Opteron processors, 12 cores each
  - 2 NUMA nodes per socket
- 20xNUMA4 (Grid'5000)
  - 2 AMD Opteron processors, 2 cores each
  - 20 machines on a cluster

**Load Balancers** 

#### State-of-the-Art Load Balancers

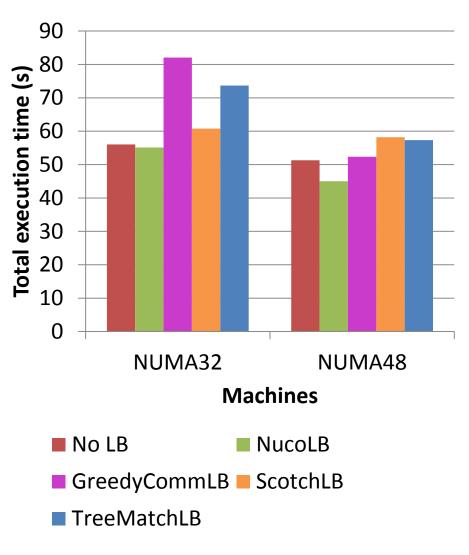
- Used for performance comparisons
- Available with Charm++
- GreedyCommLB: greedy load balancer
- ScotchLB: load balancer based on graph partitioning
- TreeMatchLB: load balancer that maps the communication graph to the machine topology graph
- All consider the load and communication behaviour of the tasks

- 2 benchmarks
  - kNeighbor
  - stencil4d
- 1 application
  - LeanMD

- Average execution time
  - Minimum of 20 executions
  - 95% of statistical confidence
    - Student t-distribution
  - 5% relative error

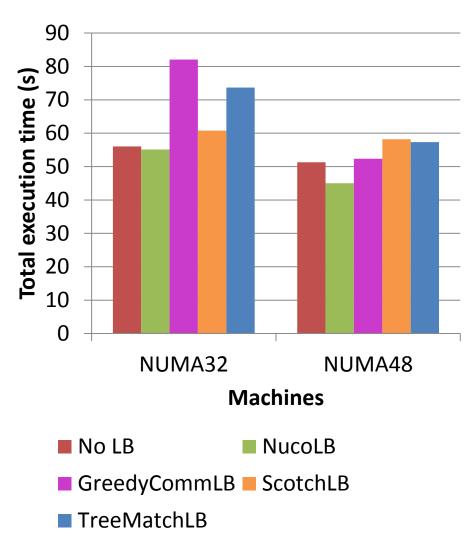
**Benchmark Results** 

- kNeighbor
  - Each task
    communicates with k
    other (k=7)
  - Communication-bound
  - 400 tasks, 8 KB
    messages, 50 iterations
  - Regular, load balanced
  - One load balancing call at each 10 iterations

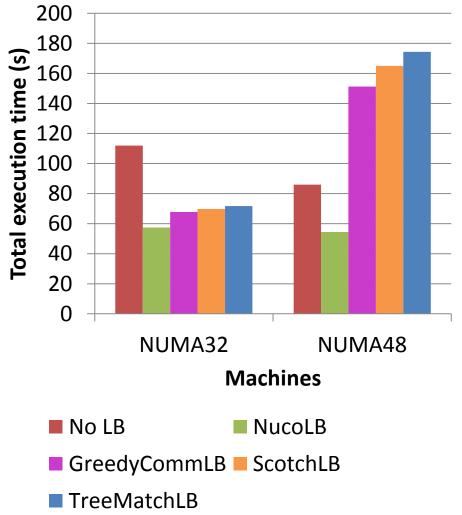


#### **Benchmark Results**

- kNeighbor
  - Low load balancing overhead
    - 150 ms per LB call, other take 700 ms
  - Speedup of 1.14 over
    no LB on NUMA48
    - Reduces the costs of communication
    - 6% load reduction



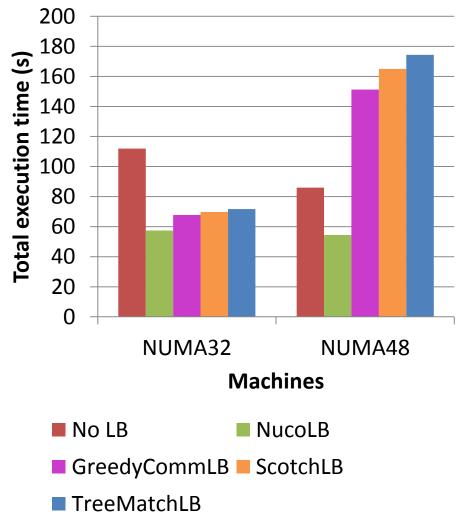
**Benchmark Results** 



### stencil4d

- Four dimensional stencil
- 256 tasks, 50 iterations
  - 4x4x4x4 stencil
- Compute-bound
- Initially imbalanced
- Large memory footprint
- One load balancing call at each 10 iterations

**Benchmark Results** 

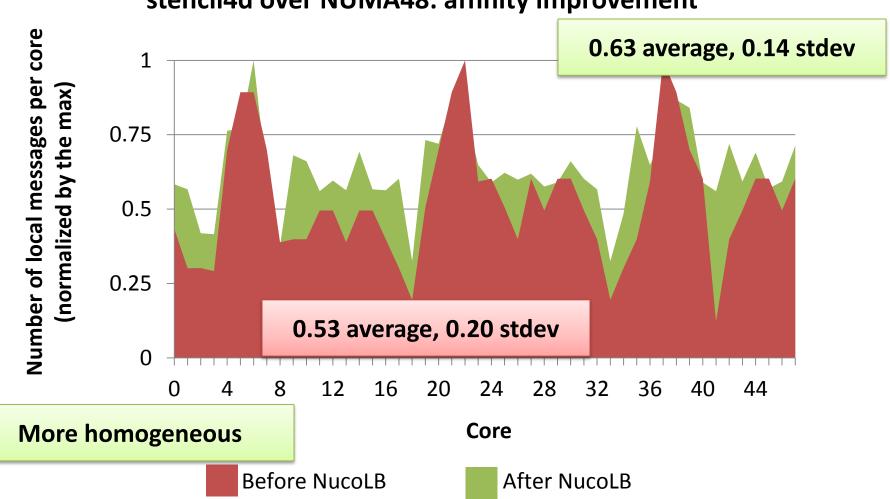


### stencil4d

- Speedup of 1.18 over the second best LB on NUMA32
- Small LB overhead
  - Does not increase cache misses and page faults
  - Other LBs increase cache misses by 31% and page faults by 19%

### – Improves affinity

**Benchmark Results** 

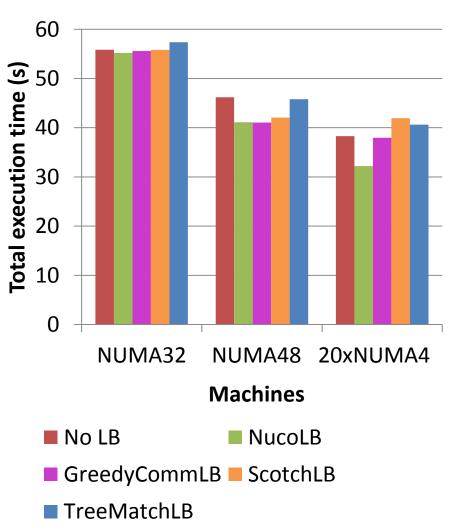


stencil4d over NUMA48: affinity improvement

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**Application Results** 

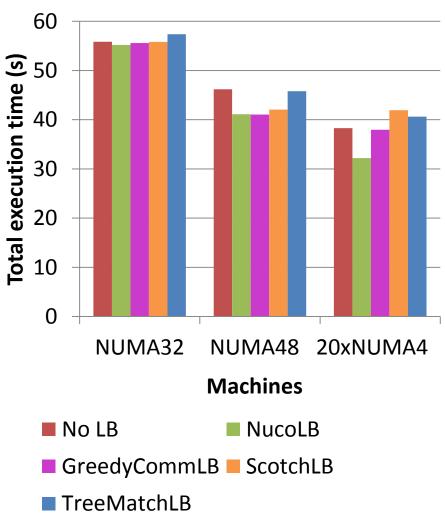
- LeanMD
  - Molecular dynamics application
  - Highly irregular
  - Compute-bound
  - 1875 tasks, 300
    iterations
  - One load balancing call at each 60 iterations



**Application Results** 

### LeanMD

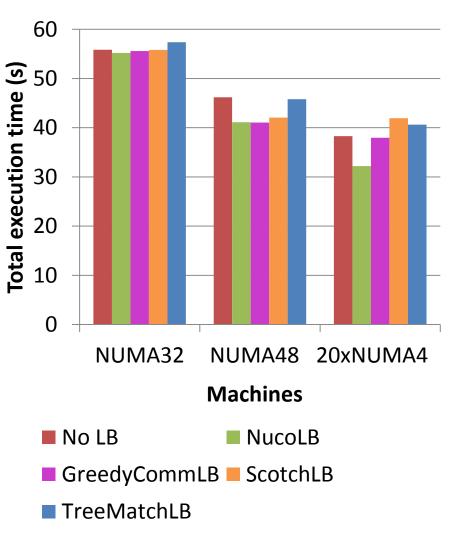
- No changes on NUMA32
  - 95% core usage from the start
  - Too many tasks per core
- Speedup of 1.12 over the baseline on NUMA48
  - Negligible overhead to move tasks around



**Application Results** 

### LeanMD

- 19% reduction on total execution time on 20xNUMA4
  - 93% core usage
  - Iteration time reduction from 114 ms to 90 ms
  - 11 to 18 times less migrations than other LBs
- Overloads some cores to reduce communication overhead



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Review

### Context

- Handle irregular applications over non-uniform, hierarchical architectures
- Reduce communication costs and idleness

### • Approach

- Combine information about the application behaviour and the machine topology model in load balancing algorithms
- Results
  - Low overheads, affinity and efficiency improvements
  - Performance portability over different environments

**Source Code** 

#### HieSchella Project

45

- <u>Hie</u>rarchical <u>Sche</u>duling for <u>Large Scale</u>
  <u>A</u>rchitectures
- Load balancers source code
- <u>https://forge.imag.fr/projects/hieschella/</u>

**Publication** 

Presentation on MMMM 2012, London

– Memory Management for Many- and Multicore

• Paper accepted on ICPP 2012



- A Hierarchical Approach for Load Balancing on Parallel Multi-core Systems
- Poster on ISC 2012

**Future Work** 

### Current and Future Work

- Improve/extend other load balancers
  - MultiCoreLB, TopoAwareLB
- Extend hwloc to provide communication costs
  - hwloc: Portable Hardware Locality library
    - Open-MPI + Inria Bordeaux
  - Latency and bandwidth
  - Easier distribution

**Future Work** 

### • Current and Future Work

- Consider additional information
  - Bandwidth
  - Contention
- Scalability, distributed algorithm
- Experiment on different platforms
  - ARM processors
- Extend algorithms for heterogeneous architectures
  - GPUs, different processing powers

# Load Balancing for Parallel Multi-core Machines with Non-Uniform Communication Costs

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# Thank you.

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