

Soft Performance Analysis for Parallel and Distributed Programs

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Talk Outline

- ❖ Motivation
- ❖ Outline of soft performance analysis approach
- ❖ Performance score and similarity measure
- ❖ Some soft analysis techniques
- ❖ Conclusion and future work

Motivation

- ❖ Lack of the specification and control of inexact parameters, commands and requests in existing performance analysis tools
- ❖ Performance tools do not interact with the user through high-level notation (e.g., words)
- ❖ Graphics techniques are very useful, but not suitable for performance analysis of large-scale and complex applications



Picture taken from
a talk of D.
Kranzlmüller (Uni.
Linz)

- ❖ **Our approach:** apply soft computing, similarity measure, machine learning in performance analysis

Simple Example: Soft vs Hard Analysis

❖ Hard computing

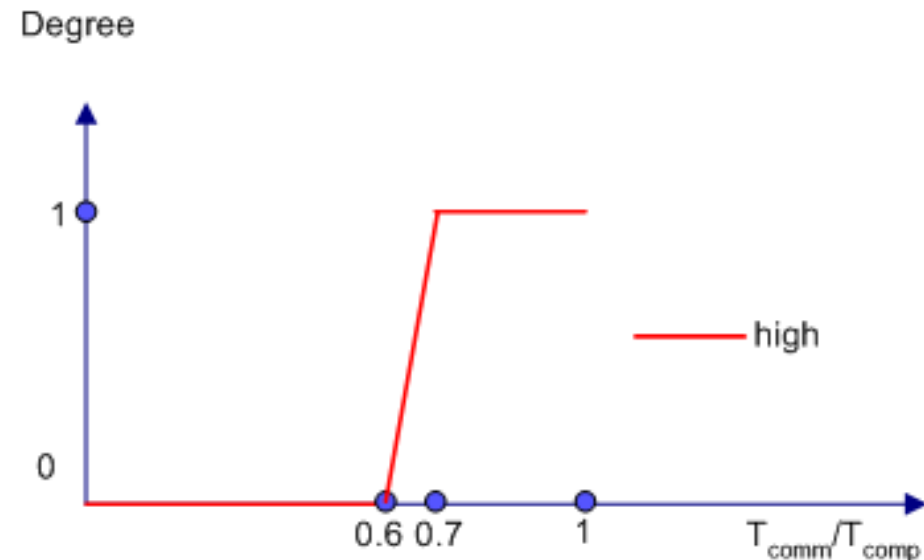
- Apply exact methods
- Binary logic, crisp system, numerical analysis

If $T_{\text{comm}}/T_{\text{comp}} > 0.7$ then r have high communication to computation ratio

❖ Soft Computing

- Support imprecision and uncertainty
- Computing with words

If ($T_{\text{comm}}/T_{\text{comp}}$ is high) then r have high communication to computation ratio with x degree



Existing works

- ❖ Fuzzy logic for performance monitoring, e.g. performance contracts (Pablo)
 - ❖ Using classification techniques based on machine learning, multivariate statistical techniques (e.g., done by Vetter and colleagues)
 - ❖ APART performance property characterizes specific negative performance behavior of code regions
 - ❖ Recent work applying data clustering in TAU (Uni. Oregon, to appear in SC05)
- ⇒ fuzzy logic has not been exploited in data analysis techniques, e.g., performance classification
- ⇒ not interact with the end user through high-level notation, e.g. linguistic query

Outline of Soft Performance Analysis Approach

- ❖ Performance values are mapped into performance scores
- ❖ Performance characteristic terms are represented by a fuzzy set
 - A set of perf. characteristic terms describes possibilities of a metric
- ❖ To analyze the performance and interpret performance results with linguistic terms
- ❖ Similarity theory and machine learning: similarities and differences among performance data items
- ❖ Focuses of this talk
 - **Conceptual framework:** How can we apply soft computing into performance analysis
 - **Interaction between performance tools and the user:** Through high level notions and concepts expressed in linguistic expressions
 - **Potential applications of soft performance analysis**

Preliminaries

- ❖ Performance data
 - A program consists of a set of (instrumented) code regions
 - Each code region is measured with a set of n metrics
- ❖ Performance experiment data used obtained from
 - 3DPIC, an MPI program, simulates the interaction of high intensity ultrashort laser pulses with plasma in three dimensional geometry
 - LAPW0 calculates the effective potential of the Kohn-Sham Eigen-Value problem, implemented in Fortran MPI
 - Stommel, OpenMP/MPI program, solves the 2d Stommel Model of Ocean Circulation using a Five-point stencil and Jacobi iteration.

Performance Score

❖ Performance score concept

- Map a value of metric m , v , into $[0,1]$. Performance score, s , of v is defined by

$$s = \mathcal{O}(v), \mathcal{O}(v):[0,V] \rightarrow [0,1]$$

- $\mathcal{O}(v)$ is the membership function, V is the maximum value of m obtained from the base.
- Each code region is represented by a vector of scores
- Overall weighted average (OWA) for performance scores

$$\text{OWA}(\vec{s}) = \frac{\sum_{i=1}^n (s_i * w_i)}{\sum_{i=1}^n w_i}$$

Performance Score (cont.)

- ❖ The base is dependent on the scope of the analysis
 - Analysis can be done within a code region, a thread or the entire program
- ❖ $[0, 1]$: 0 means lowest score, 1 means highest score
 - Semantics is defined by specific implementations
- ❖ Membership functions are also analysis-dependent
 - Examples: linear, S-function, etc.
- ❖ Performance score concept allows to normalize performance metrics but considering
 - The dynamics and flexibility
 - The uncertainty and imprecision
- ❖ Used in dynamic tuning, ranking, clustering, etc.

Ranking Analysis

- ❖ Widely used in distinguishing significant and insignificant components
 - Which child code regions of a code region have strong impact on the performance of the parent?
- ❖ Ranking based on raw measurement value is difficult to interpret and compare the significance of the performance

The image displays two side-by-side screenshots of a performance analysis tool, likely Visual Studio's Performance Analyzer, showing code region analysis for a process named 'gsr403'.

The left window, titled "Performance Rank based on", shows a tree view of code regions. The root is "gsr403", followed by "Process 0" and "Thread 0". Under "Thread 0", there is a folder for "Region 1:RELATIVISTIC_PLASMA_PROPAGATION[CR_P:1:186]". Inside this region, several child regions are listed with their ranks:

- Region 7:MAIN[CR_S:173:183](rank=0.999993531881636)
- Region 46:IONIZE_MOVE[CR_A:1281:1788](rank=0.4516)
- Region 26:SR_E_FIELD[CR_A:653:723](rank=0.0021296)
- Region 48:CAL_POWER[CR_A:2244:2323](rank=0.00198)
- Region 13:MPI_RECV[CR_MPIRECV:1132:1134](rank=6.0)
- Region 52:MPI_RECV[CR_MPIRECV:1714:1716](rank=6.0)
- Region 15:MPI_RECV[CR_MPIRECV:1155:1157](rank=5.0)
- Region 64:MPI_RECV[CR_MPIRECV:1846:1848](rank=4.0)
- Region 17:MPI_RECV[CR_MPIRECV:1179:1180](rank=1.0)

The right window, titled "Performance Rank based on Fuzzy", shows a similar tree view. The root is "gsr403", followed by "Process 0" and "Thread 0". Under "Thread 0", there is a folder for "Region 1:RELATIVISTIC_PLASMA_PROPAGATION[CR_P:1:186]". Inside this region, several child regions are listed with their ranks:

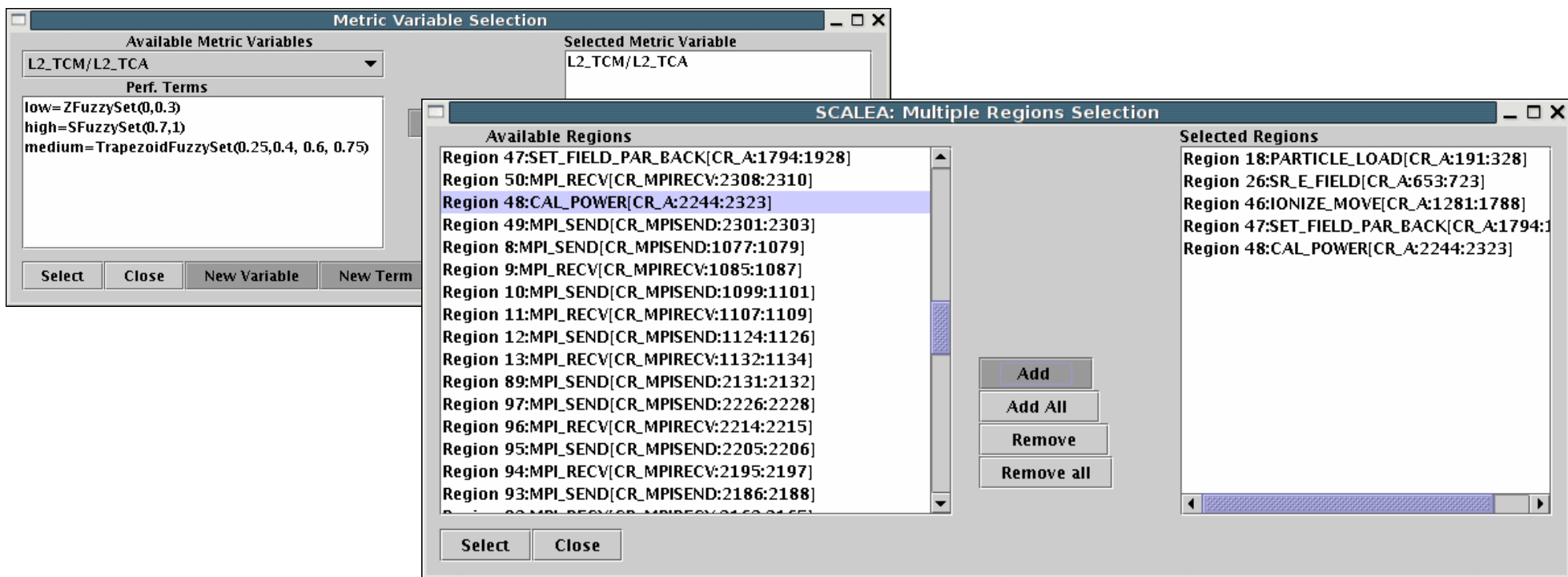
- Region 7:MAIN[CR_S:173:183](rank=1.0)
- Region 57:MPI_SEND[CR_MPISEND:1770:1772](rank=0.07648119182319221)
- Region 69:MPI_SEND[CR_MPISEND:1904:1906](rank=0.06450781781946)
- Region 8:MPI_SEND[CR_MPISEND:1077:1079](rank=0.04915781679412661)
- Region 26:SR_E_FIELD[CR_A:653:723](rank=0.047365576887277515)
- Region 76:MPI_RECV[CR_MPIRECV:1983:1984](rank=0.03992459719258101)
- Region 10:MPI_SEND[CR_MPISEND:1099:1101](rank=0.026263382338476103)
- Region 15:MPI_RECV[CR_MPIRECV:1155:1157](rank=0.025321410081430525)
- Region 52:MPI_RECV[CR_MPIRECV:1714:1716](rank=0.01578457339660404)
- Region 13:MPI_RECV[CR_MPIRECV:1132:1134](rank=0.015350390589907456)
- Region 64:MPI_RECV[CR_MPIRECV:1846:1848](rank=0.014057219476519379)
- Region 93:MPI_SEND[CR_MPISEND:2186:2188](rank=0.006558006237502663)
- Region 48:CAL_POWER[CR_A:2244:2323](rank=0.0061552660484693305)

Fuzzy-based Performance Classification

1. Define a set of *performance characteristic terms* T for a given metric

$$T = \{t_1, t_2, \dots, t_n\}$$

2. A term is represented by a fuzzy set
3. Performance data are classified according to terms



Fuzzy-based Performance Classification (cont.)

Code Region	Classes
Region 18:PARTICLE_LOAD[CR_A:191:328]	high(degree=0.937)/
Region 26:SR_E_FIELD[CR_A:653:723]	medium(degree=1)/
Region 47:SET_FIELD_PAR_BACK[CR_A:1794:1928]	medium(degree=0.549)/
Region 48:CAL_POWER[CR_A:2244:2323]	low(degree=0.007)/medium(degree=0.28)/

SCALEA: Profile/Trace Analysis

File View Code Region Analysis

DRG Profile

- Experiment Experiment
- gsr403
- Process 0
- Thread 0
 - Region 1:RELATIVISTIC_PLASMA_PROPAGATION[CR_P:1:1]
 - Region 2:MPI_INIT[CR_MPISTARTUP:23:23]
 - Region 3:MPI_COMM_RANK[CR_MPIOTHER:24:24]
 - Region 4:MPI_COMM_SIZE[CR_MPIOTHER:25:25]
 - Region 6:PARALLEL_INIT[CR_S:27:28]
 - Region 7:MAIN[CR_S:173:183]
 - Region 18:PARTICLE_LOAD[CR_A:191:328]
 - Region 20:MPI_RECV[CR_MPIRECV:317:318]
 - Region 26:SR_E_FIELD[CR_A:653:723]
 - Region 8:MPI_SEND[CR_MPISEND:1077:1079]
 - Region 10:MPI_SEND[CR_MPISEND:1099:1101]
 - Region 13:MPI_RECV[CR_MPIRECV:1132:1134]
 - Region 15:MPI_RECV[CR_MPIRECV:1155:1157]
 - Region 17:MPI_RECV[CR_MPIRECV:1179:1180]
 - Region 46:IONIZE_MOVE[CR_A:1281:1788]
 - Region 47:SET_FIELD_PAR_BACK[CR_A:1794:1928]
 - Region 52:MPI_RECV[CR_MPIRECV:1714:1716]
 - Region 54:MPI_RECV[CR_MPIRECV:1730:1731]
 - Region 57:MPI_SEND[CR_MPISEND:1770:1772]
 - Region 59:MPI_SEND[CR_MPISEND:1787:1788]
 - Region 64:MPI_RECV[CR_MPIRECV:1846:1848]

```
! Initializing for a new run end
-----!
ISIS$ CR PARTICLE_LOAD, CR_MPISEND, CR_MPIRECV BEGIN
-----!
! Particle loader for a new run begin
-----!

if (myid.eq.0) then
  print *, 'Loading particles ...'
end if

! Test begin :
! goto 1234
! Test end :

i=1
DO iz=1,gas_cells_z*ppc_z
DO iy=1,gas_cells_y*ppc_y
DO ix=1,gas_cells_x*ppc_x

if (.gt.qq_array) then
  print *, 'qq_max =',
& gas_cells_z*gas_cells_y*gas_cells_x
& ppc_x*ppc_y*ppc_z,
& ' > qq_array =', qq_array
  PAUSE 'Warning !!!!!!!!!!!!!!!!!!!!!!!'
end if

qq(i,1)=myid

ppc(3)=hx*(ix-1)/npx*x*1.0d0+gas_begin_x
```

Metric Name	Metric Value
odata_rcv	920

Fuzzy-based Performance Search

- ❖ Existing performance tools
 - Do not offer the possibility of search performance data with linguistic query
- ❖ PERFORMANCE Query Language based on fuzzy logic (PERFQL)
 - Performance search based on linguistic expressions
 - Easily to define/understand queries

```
<PERFQL_Statement> ::= <PERFQL_Expr> | <PERFQL_Statement>  
                        OR <PERFQL_Expr>  
<PERFQL_Expr> ::= <PERFQL_Term> | <PERFQL_Expr>  
                        AND <PERFQL_Term>  
<PERFQL_Term> ::= (<METRIC_Expr> is <F_Expr>)
```

Metric or Metric Expression

wtime

L2_TCM/L2_TCA

odata_send/wtime

Fuzzy Expression

HIGH_EXECUTION_TIME

very HIGH_EXECUTION_TIME

slightly POOR_SEND_OVERHEAD

Fuzzy-based Performance Search (cont.)

The image shows a screenshot of a performance analysis tool interface. It features several windows:

- PERFQL Input:** Shows available metric variables (wtime) and performance terms (HIGH_EXECUTION_TIME=LeftLinearFuz).
- SCALEA: Profile/Trace Analysis:** Displays a DRG Profile tree with various code regions. Region 46 (IONIZE_MOVE) is highlighted.
- Code Region Analysis:** Shows a detailed view of the selected region, including code snippets and a table of metrics.

Two callout boxes provide additional context:

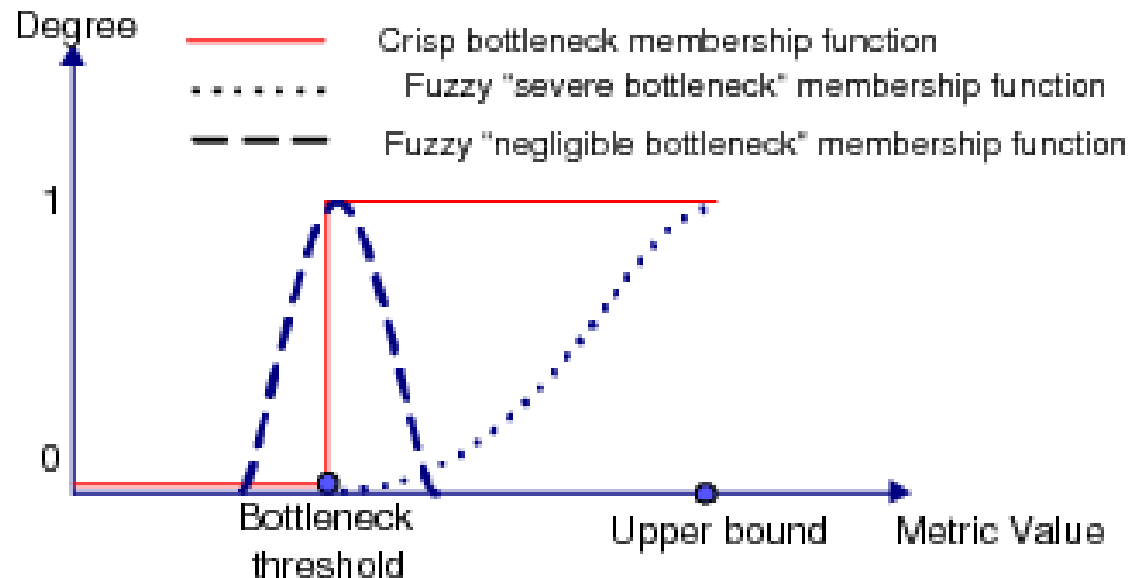
- Yellow Callout (top left):** "Assume any code region takes more than 20% total execution is HIGH_EXECUTION_TIME".
- Yellow Callout (bottom right):** "New query with cache misses condition".

The metrics table in the bottom right window is as follows:

Metric Name	Metric Value
nsubs	0
ncalls	800
L2_TCA	1,079,784,122
wtime	115,880,587
L2_TCM	118,057,448
stime	170.000

Fuzzy Approach to Bottleneck Search

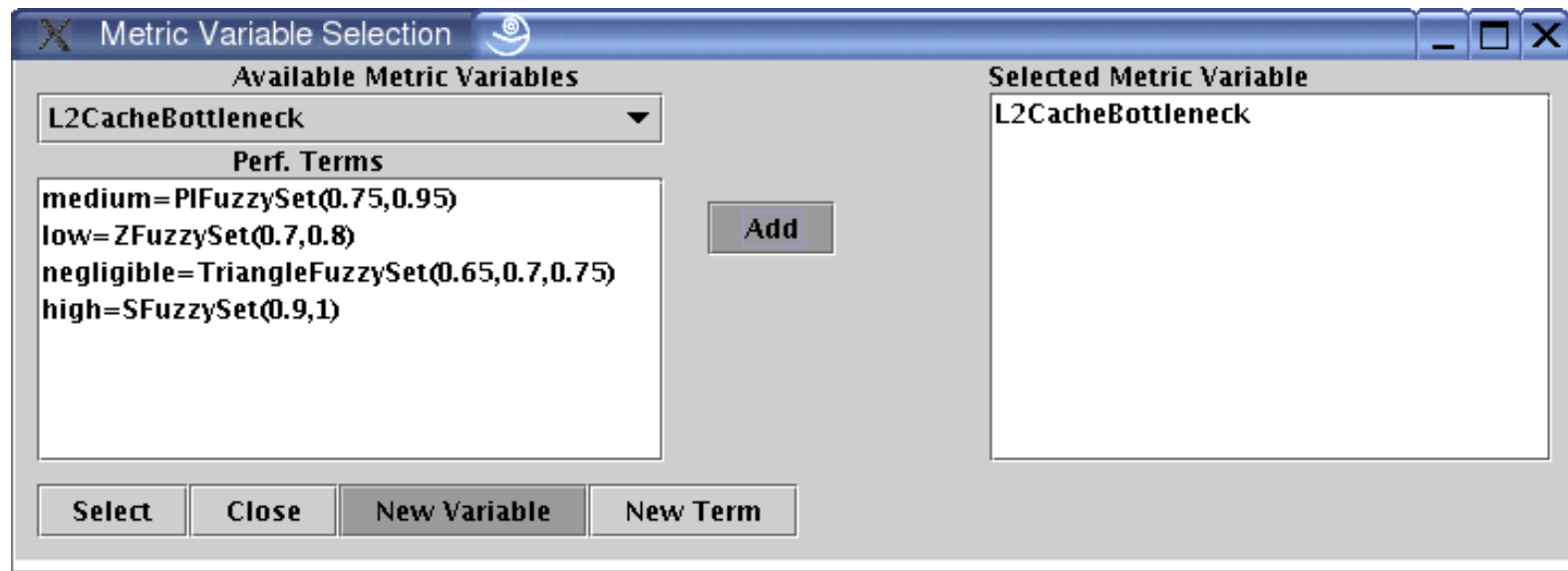
1. Using fuzzy sets to represent *bottleneck conditions*
2. Using fuzzy sets to represent *negligible bottlenecks*



❖ Search results

- Indicate the *degree of bottleneck*
 - We can use the degree of bottleneck for further tasks
- Locate *negligible bottlenecks*
 - We may not find any bottlenecks because the condition is not exact

Bottleneck Search: Simple Example



❖ Search for low, medium and high *degree of bottleneck*

Fuzzy-based Search		
Code Region	L2_TCM/L2_TCA	Bottleneck
PARTICLE_LOAD	0.9497242344418835	Medium (degree=0.502)/High (degree=0.495)/

❖ Search also *negligible bottlenecks*

Fuzzy-based Search		
Code Region	L2_TCM/L2_TCA	Bottleneck
MPI_SEND	0.6550124668722567	Negligible (degree=0.1)/
PARTICLE_LOAD	0.9497242344418835	Medium (degree=0.502)/High (degree=0.495)/

Performance Similarity Measure

- ❖ Problems:
 - Difficult to observe and perceive the performance similarity and difference through complex visualization
- ❖ Performance similarity measure indicates the performance similarity among code regions and among experiment factors

$$\text{sim}(o_i, o_j) \rightarrow [0, 1]$$

- 0 denotes complete dissimilarity and 1 denotes complete similarity

Performance Similarity Measure

- ❖ Performance similarity measure for code regions
 1. Using performance score concept to determine **performance scores** of region summaries rs_i and rs_j . Each rs is represented as a vector of n performance scores
 2. Determining distance measure between rs_i and rs_j . For example,

$$d_{ij} = \sqrt{\sum_{l=1}^n (s_{il} - s_{jl})^2}$$

3. Determining performance similarity between two code regions

$$\text{sim}_{ij}(rs_i, rs_j) = 1 - d_{ij}$$

Performance Similarity Analysis (cont.)

❖ Stommel:

- Similarity measure for cache accesses of Stommel application

ProcessingUnit	gsr415->0->0	gsr415->0->1	gsr415->0->2	gsr415->0->3	gsr411->1->0	gsr411->1->1	gsr411->1->2	gsr411->1->3
gsr415->0->0	1	0.944	0.659	0.659	0.893	0.893	0.659	0.659
gsr415->0->1	0.944	1	0.672	0.672	0.837	0.949	0.672	0.672
gsr415->0->2	0.659	0.672	1	1	0.602	0.683	1	1
gsr415->0->3	0.659	0.672	1	1	0.602	0.683	1	1
gsr411->1->0	0.893	0.837	0.602	0.602	1	0.786	0.602	0.602
gsr411->1->1	0.893	0.949	0.683	0.683	0.786	1	0.683	0.683
gsr411->1->2	0.659	0.672	1	1	0.602	0.683	1	1
gsr411->1->3	0.659	0.672	1	1	0.602	0.683	1	1

❖ LAPW0:

- Similarity measure based on wallclock time

CodeRegion/Experiment	2Nx4P,P4,36	2Nx4P,GM,36	3Nx4P,P4,36	3Nx4P,GM,36	3Nx4P,P4,72	3Nx4P,GM,72
Region 2:CA_MULTIPOLMENTS[CR_A:256:506]	1	0.996	0.638	0.635	0.625	0.625
Region 3:CA_COULOMB_INTERSTITIAL_POTENTIAL[CR_A:536:565]	1	0.986	0.629	0.636	0.597	0.597
Region 4:CAL_COULOMB_RMT[CR_A:635:668]	1	0.999	0.63	0.631	0.597	0.597
Region 5:CAL_CP_INSIDE_SPHERES[CR_A:678:772]	1	0.982	0.632	0.639	0.598	0.598
Region 6:FFT_REAN0[CR_OTHERSEQ:881:883]	1	0.997	1	0.997	0.981	0.981
Region 7:FFT_REAN3[CR_OTHERSEQ:889:891]	1	0.999	1	1	0.536	0.756
Region 9:FFT_REAN4_CR[CR_OTHERSEQ:915:917]	1	0.993	1	1	0.492	0.479

Performance Similarity Analysis (cont.)

❖ Performance similarity measure for experiment factors

Given a set of controllable factors $F=\{f_1, f_2, \dots, f_n\}$ and given experiments e_i and e_j

1. Factor f is described by a membership function
2. Determine similarity measure between f of e_i and e_j , $\text{sim}_f(e_i, e_j)$
3. Analyze relations among similarity measures for code regions and experiment factors

❖ LAPW0

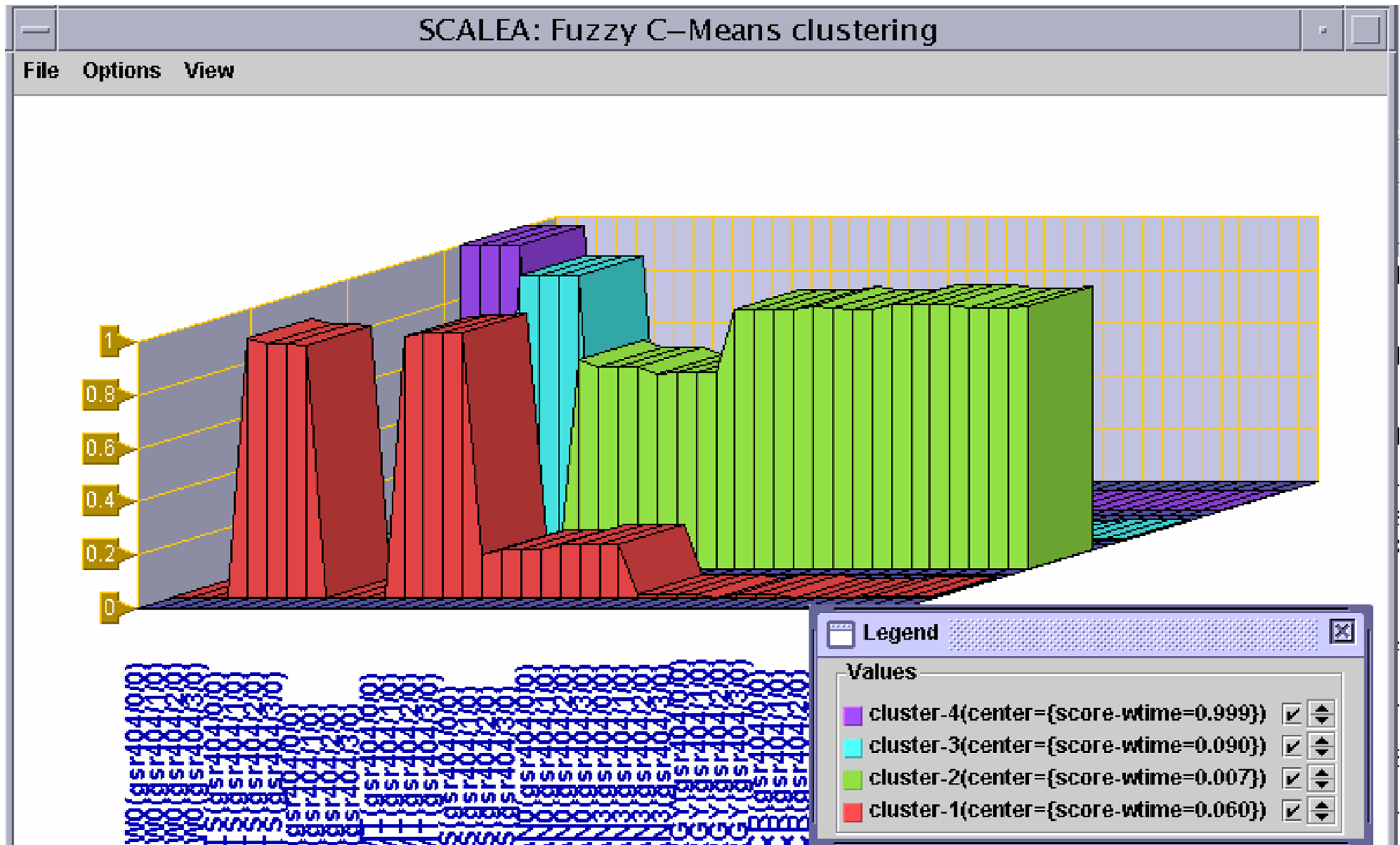
Factor	Fuzzy Set	Range	Factor Category
atoms	linear	[0,72]	problem size
CPU	S-function	[0,64]	machine
network	S-function	[0,158.20]	communication

Experiments	2Nx4P, P4,36	2Nx4P, GM,36	3Nx4P, P4,36	3Nx4P, GM,36	3Nx4P, P4,72	3Nx4P, GM,72
$\text{sim}_{f_{atoms}}(\{(atoms,1)\})$	1	1	1	1	0.5	0.5
$\text{sim}_{f_{CPU}}(\{(CPU,1)\})$	1	1	0.9531	0.9531	0.9531	0.9531
$\text{sim}_{f_{network}}(\{(network,1)\})$	1	0.1519	1	0.1519	1	0.1519
$\text{sim}_o(\{(wtime,1)\})$	1	0.996	0.638	0.635	0.625	0.625

Similarity analysis for
CA_MUTIPOLMENTS
region

Fuzzy C-Means Clustering

❖ 3D PIC executed with 4 processes



Other potential applications of soft performance analysis techniques

- ❖ Decision making in dynamic performance tuning
 - Dynamic performance tuning tools: **MATE** (UAB), **Active Harmony** (J. Hollingsworth)
 - Automatically replacing components, selecting different implementations based on performance scores and performance similarity measures
- ❖ Performance data collection/reduction
 - Rules based on crisp-condition can be replaced by fuzzy rules based on performance scores
- ❖ etc.

Conclusion and Future Work

- ❖ Contributions: we proposed the soft performance analysis approach
 - Provide flexible, scalable techniques for analyzing and comparing the performance of parallel and distributed applications
 - Interact with the user through high-level notation
 - Aim to support automatic performance analysis
- ❖ However, soft performance analysis is just at an early stage
 - Not everything discussed has been fully implemented
- ❖ What should be done next
 - So far, we have just focused on conceptual framework, not on how to select membership and distance functions
 - Study the selection of membership and distance functions
 - Apply soft performance analysis for dynamic performance tuning, autonomic computing
 - Linguistic variables and fuzzy rules for SLAs (service level agreements) in the Grid