



IBM T. J. Watson Research Center

End-to-End Performance Optimization of Java Server Workloads

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- Toshio Nakatani, Kazuaki Ishizaki, Tamiya Onodera

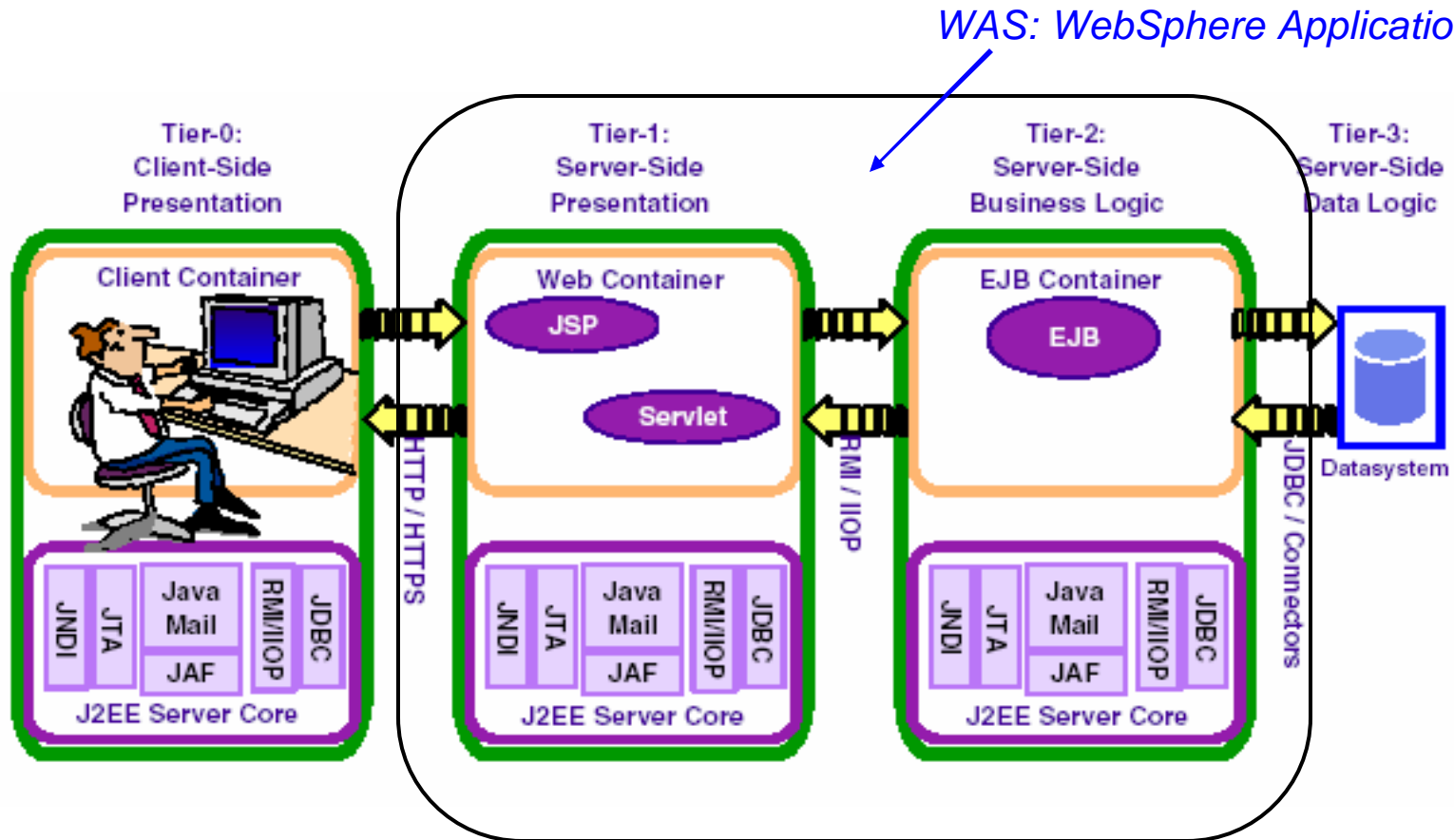
Outline

- **End-to-End Optimization Project**
 - Workload and Server Configurations
- Methodology
- Performance Characteristics
 - Method Profiling, Cache, Branch Prediction, Synchronization
- Summary

End-to-End Optimization Project

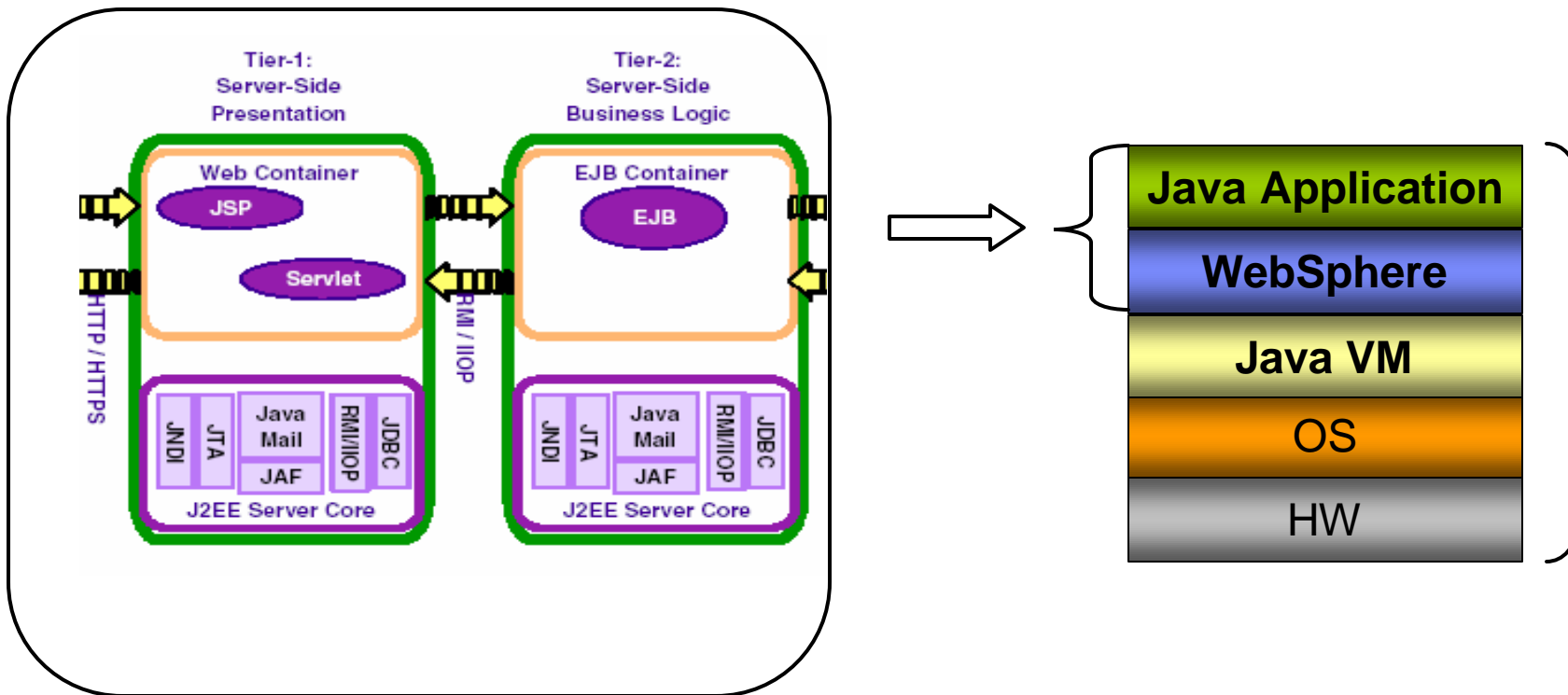
- To understand and optimize the performance of the *whole-stack, end-to-end* SW/HW layers of *commercial middleware* applications (J2EE) on IBM's current and future *high-end servers*.

Workload: J2EE Multi-tier Server w/ Application

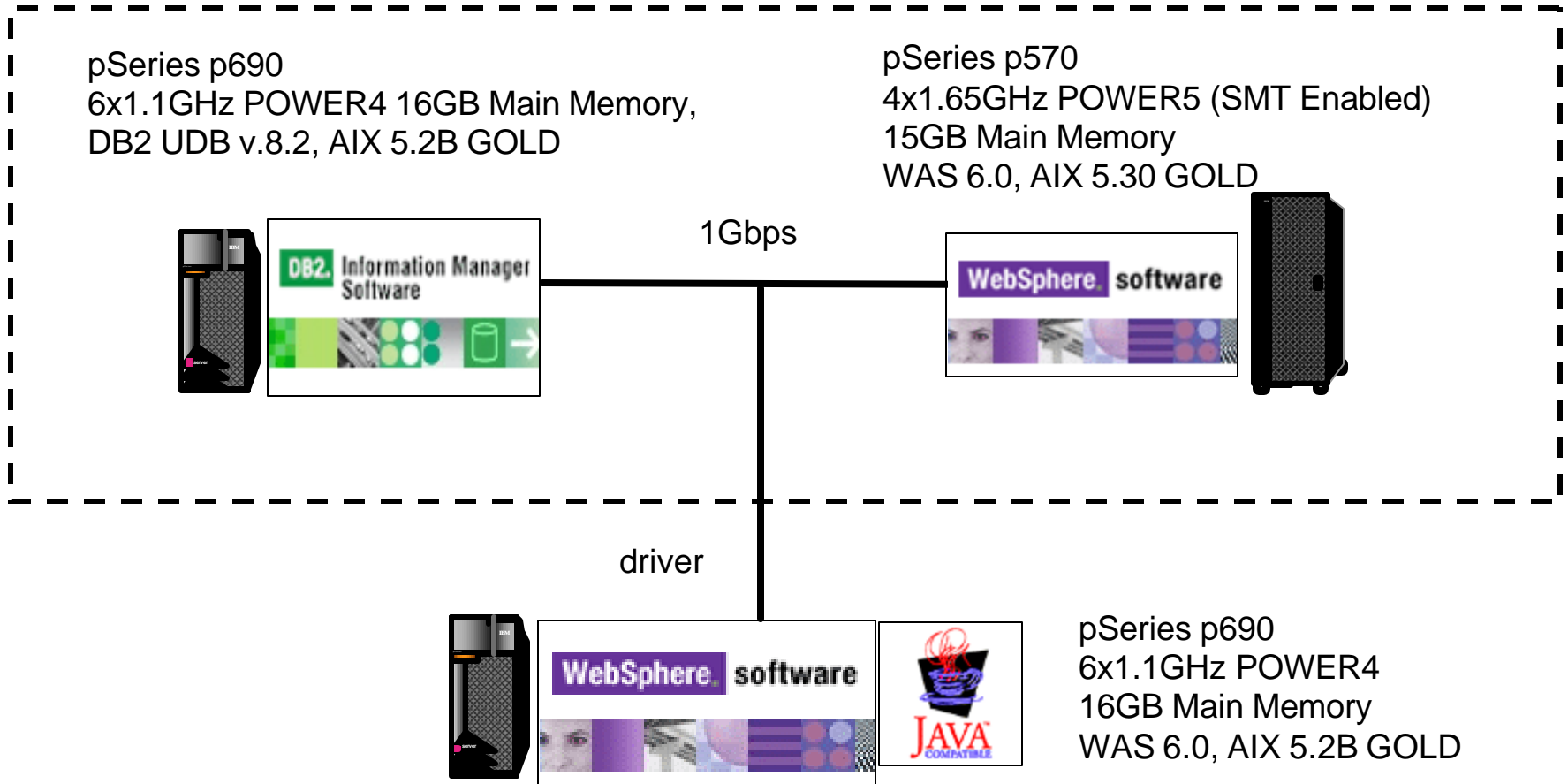


Source: Programming J2EE APIs with WebSphere Advanced by Osamu Takagiwa, et. al. Ibm.com/redbooks

J2EE Whole-Stack End-to-End Optimization



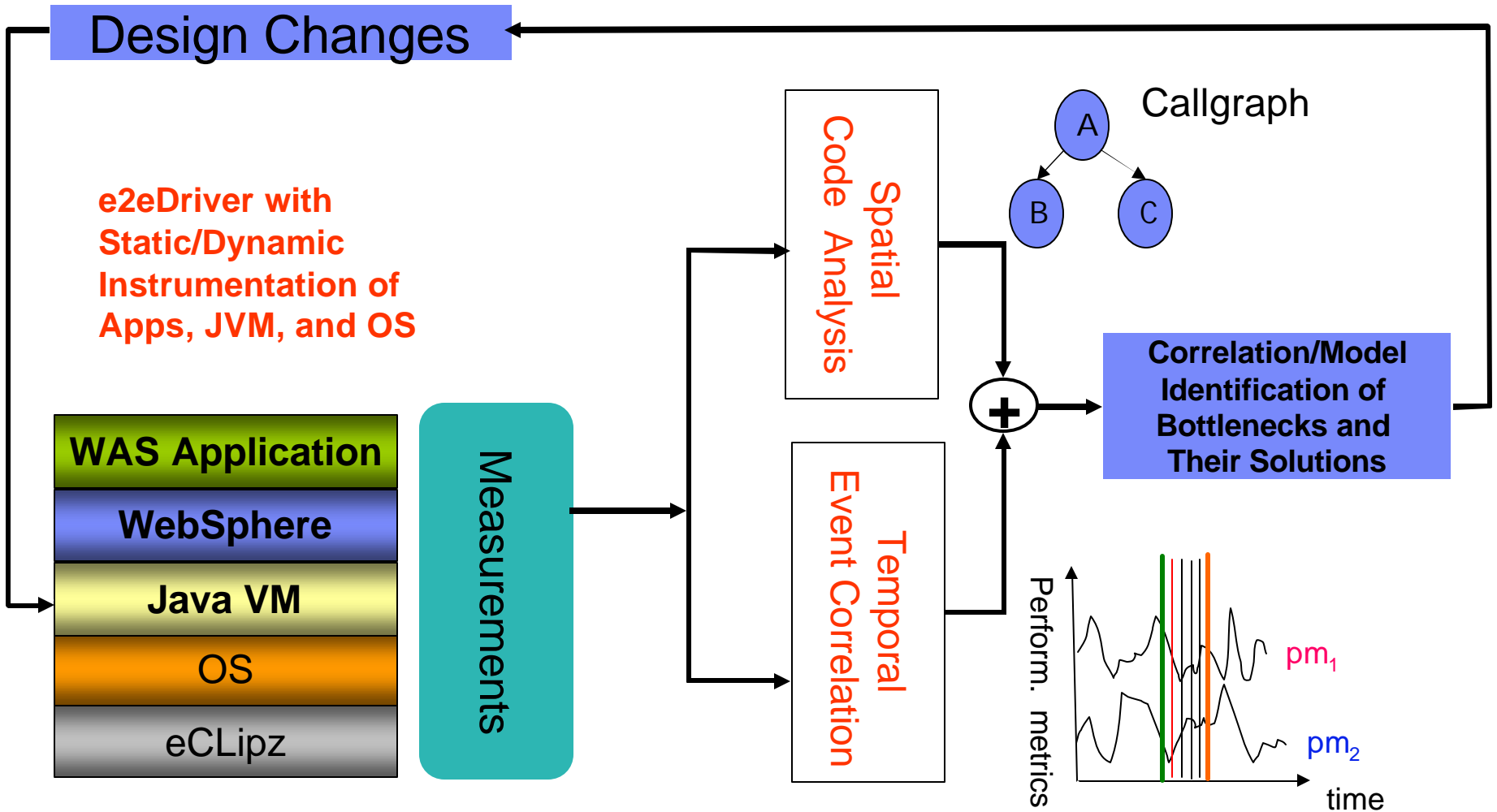
Server Configuration for SpecJAS2004/Trade6



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End-to-End Optimization Methodology

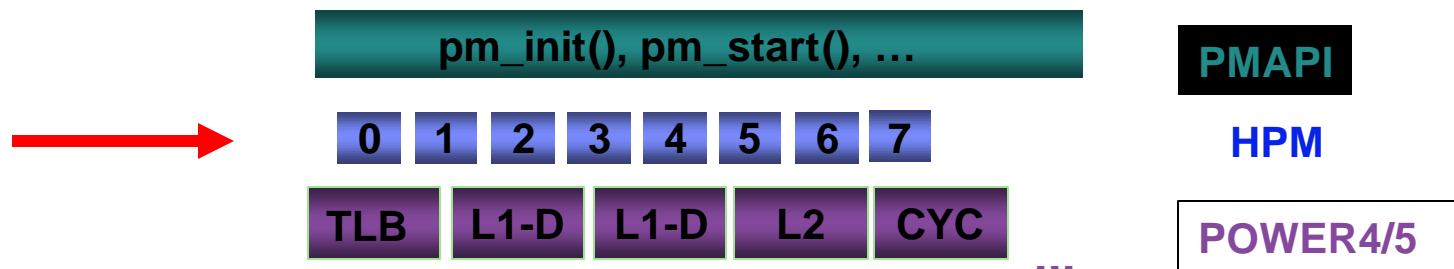


Performance Metrics and Tools

	Metrics Examples	Method
WAS Application	Response time of each transaction	Application Response Management (ARM)
WebSphere	# of executing beans, # of activated beans	Performance Monitor Infrastructure (PMI) in WAS
Java VM	# of method calls, # of GC, # of object allocations, # of syncs	Java Instrumentation
OS	# of context switches	AIX trace facility, vmstat, sar
HW	# of inst., # of loads, # of D\$ misses	HW Performance counters in POWER4/5

Hardware Performance Monitor (HPM)

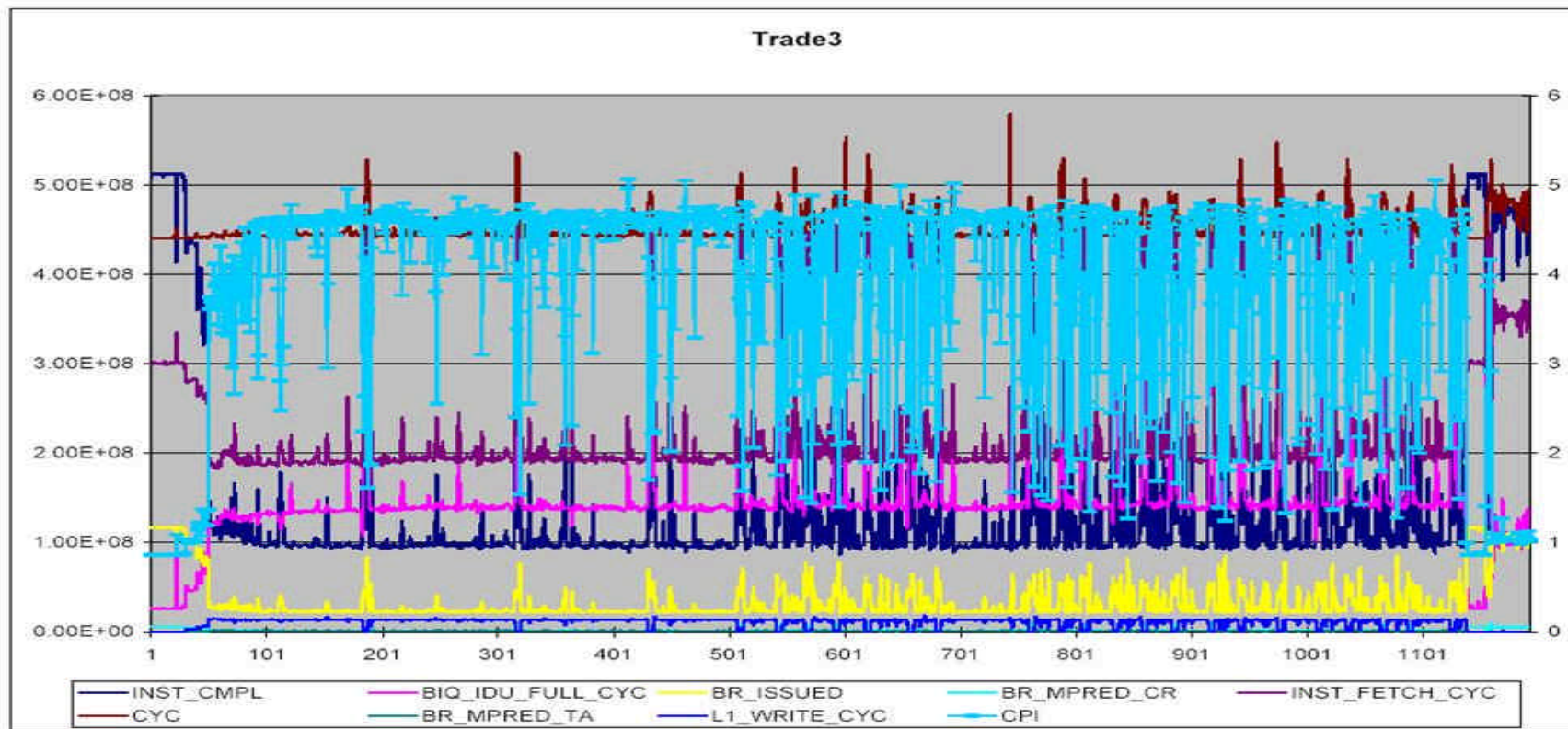
- POWER4 has 8 HPM counters that can be programmed to count HW events
 - The HW events are combined into logical *groups*
 - There are 61 groups, and 8 events per group (one event per counter)



Group 56: CPI, TLB, L1-D cache

- PM_DTLB_MISS** -- Data TLB misses
- PM_ITLB_MISS** -- Instruction TLB misses
- PM_LD_MISS_L1** - L1 D cache load misses
- PM_ST_MISS_L1** - L1 D cache store misses
- PM_CYC** ----- Processor cycles
- PM_INST_CMPL** -- Instructions completed
- PM_ST_REF_L1** -- L1 D cache store references
- PM_LD_REF_L1** -- L1 D cache load references

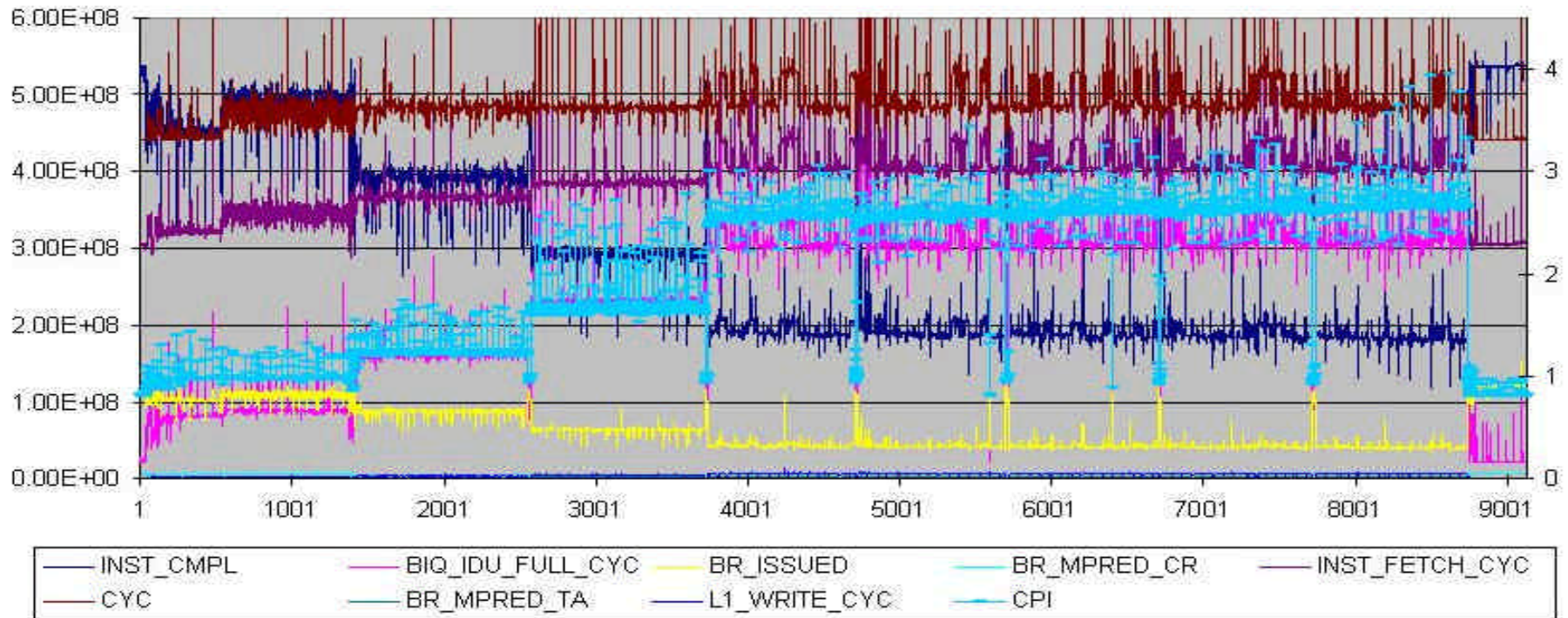
HW Performance Counters



1. Trade3/WAS/Sovereign/AIX/Power4, System+User
2. hpm counters, group 3, ~100 seconds
3. Steady state CPI = ~4.5

HW Performance Counters

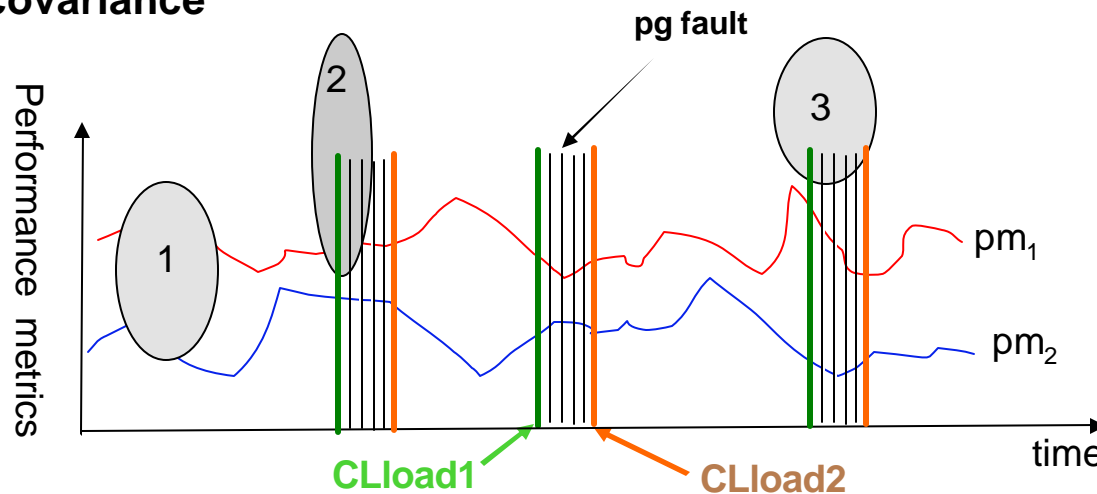
JBB/J9, System+User



1. SpecJBB/J9/AIX/Power4, System+User
2. hpm counters, group 3, ~900 seconds
3. Steady state CPI = ~2.5

Temporal Event Correlation

- 1. Micro Event \otimes Micro Event (e.g., performance metrics)**
 - CPI \otimes TLB misses
 - 2. Micro Event \otimes Macro Event**
 - TLB misses \otimes Page fault at OS
 - 3. Macro Event \otimes Macro Event**
 - Page fault at OS \otimes Class Loading
- **Temporal event correlation employs various statistical tools such as covariance**

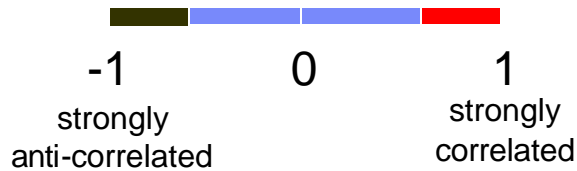


Derived Metrics - Correlations

Given two vectors $X = \{x_i\}$ and $Y = \{y_i\}$

$$\text{covar}(X, Y) = (1/n) \sum (x_i - \bar{x})(y_i - \bar{y})$$

$$\text{cc}(X, Y) = \text{covar}(X, Y) / \text{SQRT} [\text{covar}(X, X) * \text{covar}(Y, Y)]$$

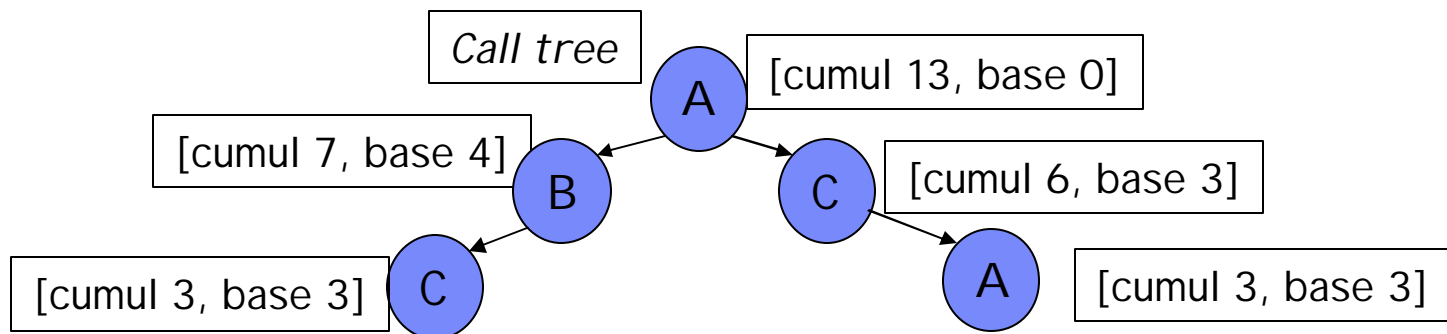
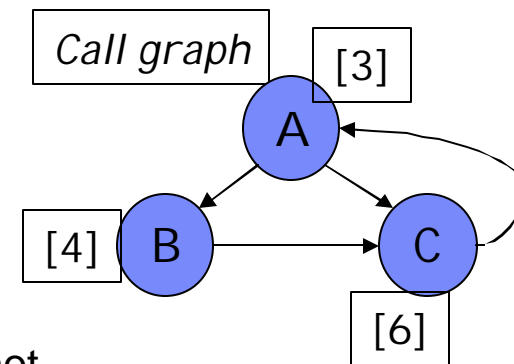


- Observe Trends: transient, steady-state, periodic
- Certain correlations are expected; spot the unexpected
- Needs systematic study

Spatial Code Analysis

Profile-based hot-code analysis

- Context-insensitive
 - Identify m hot (frequently executed) methods
 - May fail to provide the contexts in which methods are hot
- Context-sensitive
 - Identify n hot dynamic call chains
 - e.g. critical call-path information
 - May fail to recognize hot methods with uniform and low unit-cost

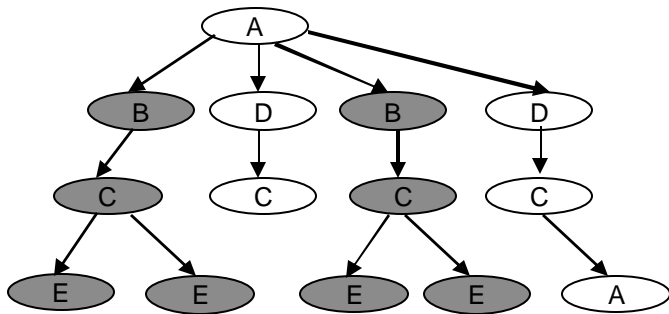


Context-Sensitive Analysis

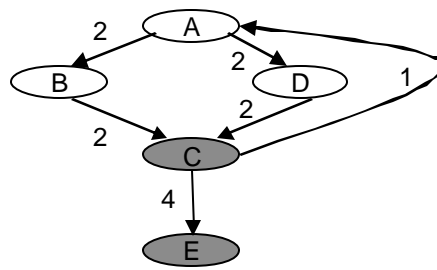
- **SPECjAppServer2004: 50% of JIT'd code execution is in 224 "hottest" methods**
 - Method profile is "flat"
 - Data profile is also "flat"
- **Profile-based hot-code analysis**
 - Identify n hot dynamic call chains
 - e.g. critical call-path information
- "Accurate, Efficient, and Adaptive Calling-Context Profiling," PLDI2006
 - X. Zhuang, M. Serrano, T. Cain, and J.-D. Choi

Context-Sensitive Analysis

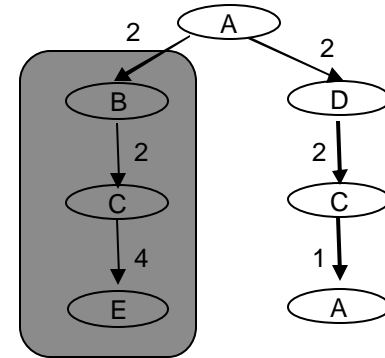
- Call sequence:
 - ‘->’: method call, ‘<-’: method return, ‘(A)’ : ‘A’ is top-of-stack
 - A -> B -> C -> E, <- , (C) ->E, < , <- , <- , (A)-> D -> C, <- . <- , (A) -> B -> C -> E , <- , (C) ->E, <- , <- , <- , (A) -> D -> C, ->A, <- , <- , <- , (A)



Call Tree



Call Graph (edge profiling)



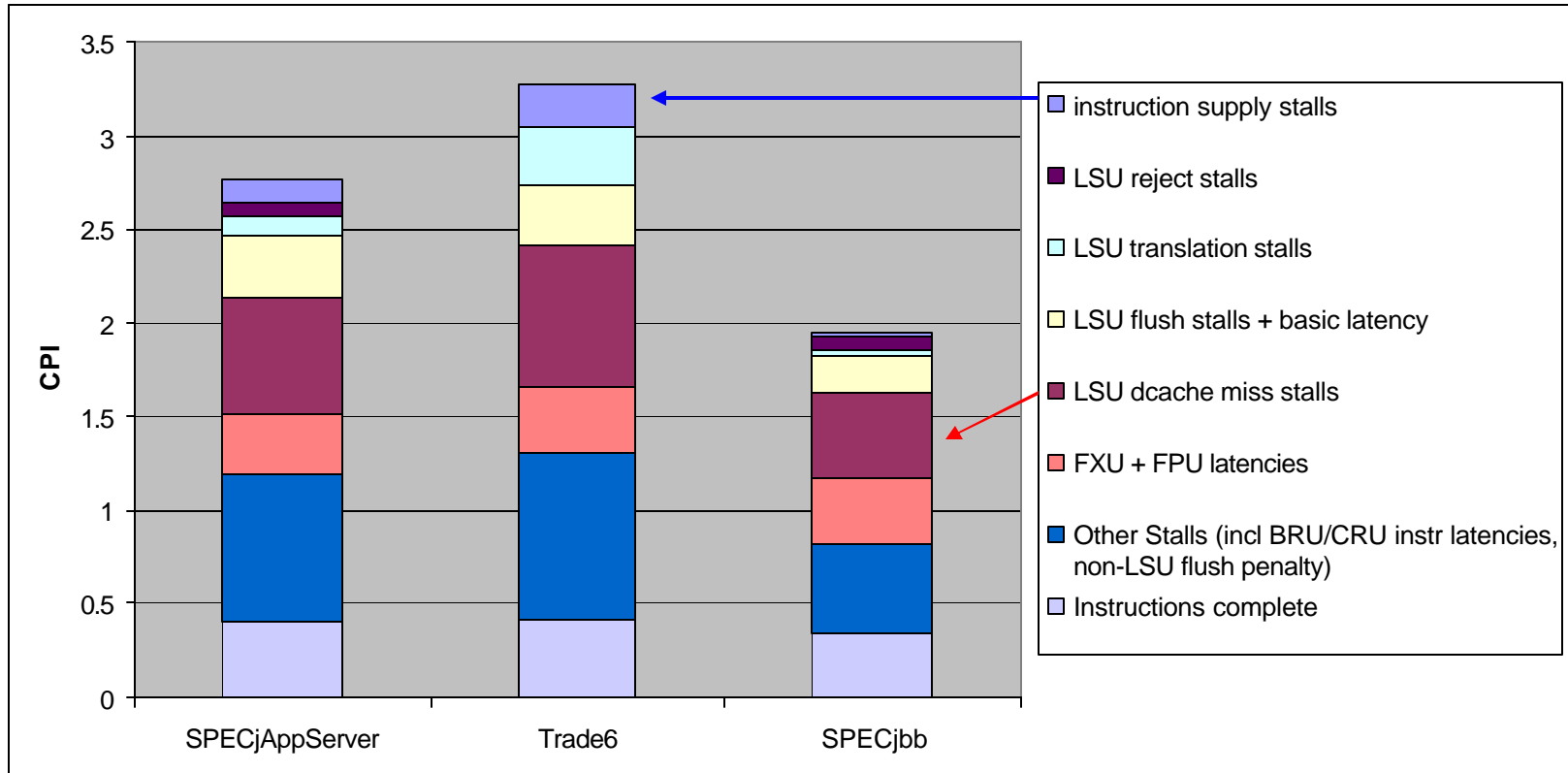
Calling-Context Tree (CCT)

- Call tree is too expensive: one node per each method call
- Call graph is too imprecise: cannot tell whether B or D is more responsible for the frequent calls of E by C
- CCT is not as expensive as call tree; on CCT it's clear B->C->E is the expensive call path.

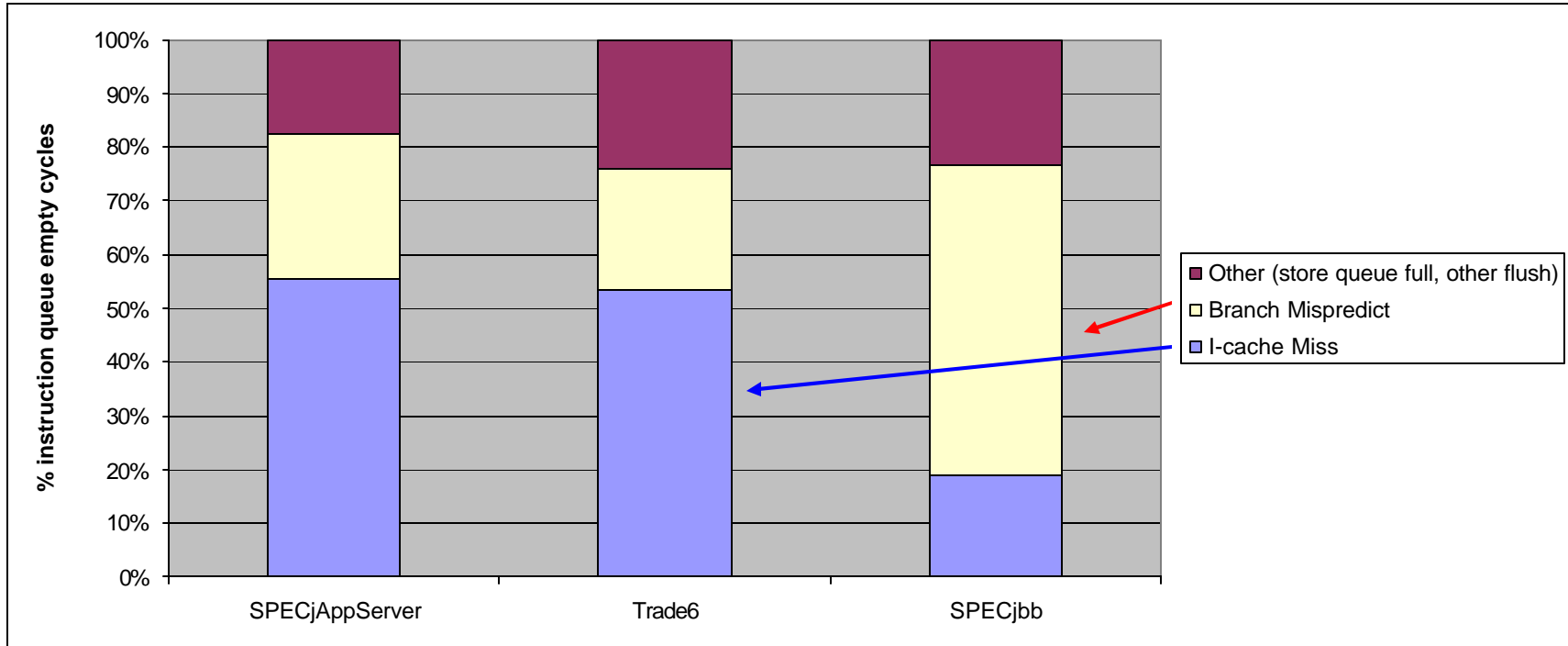
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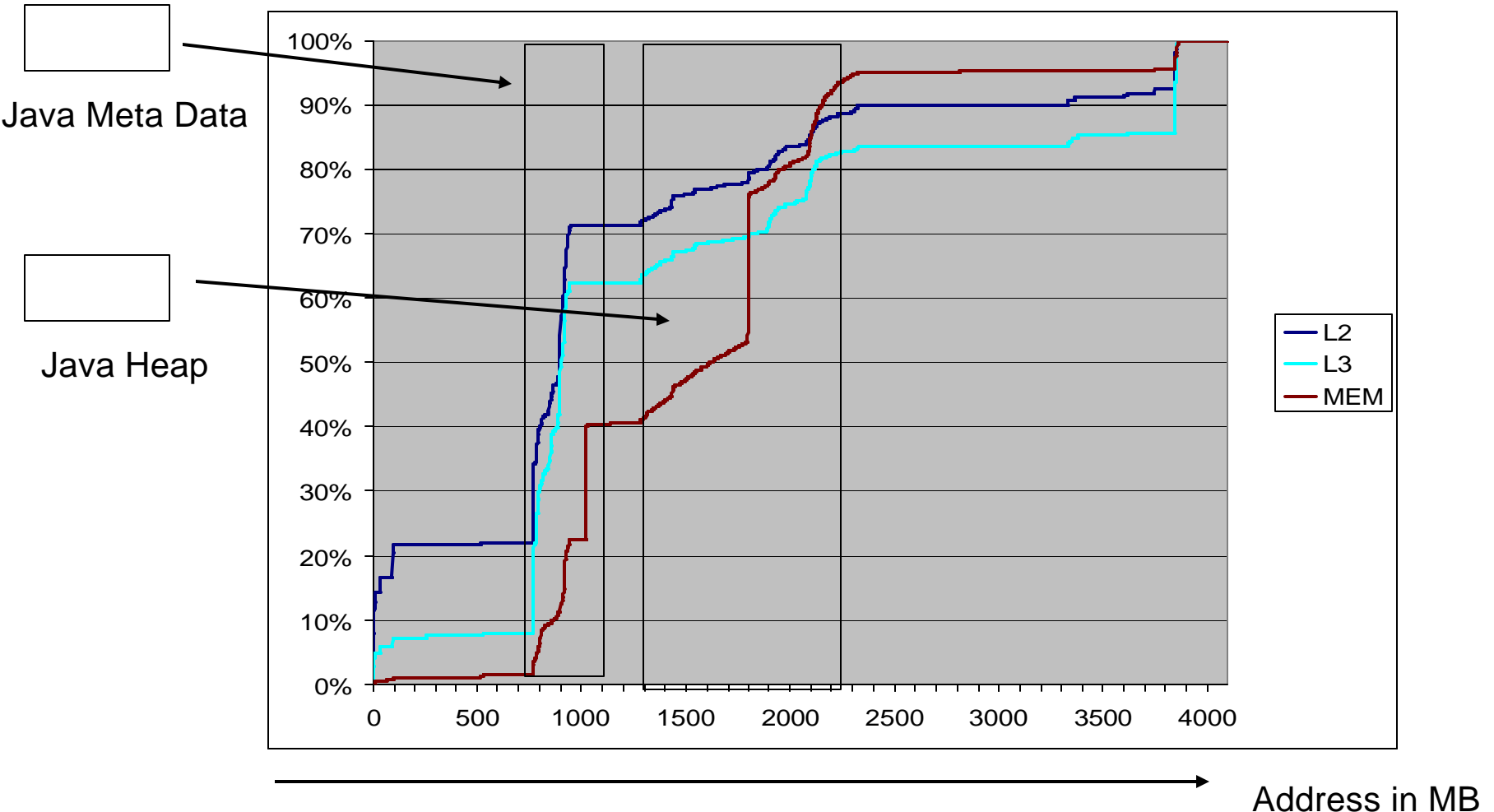
CPI Stacks



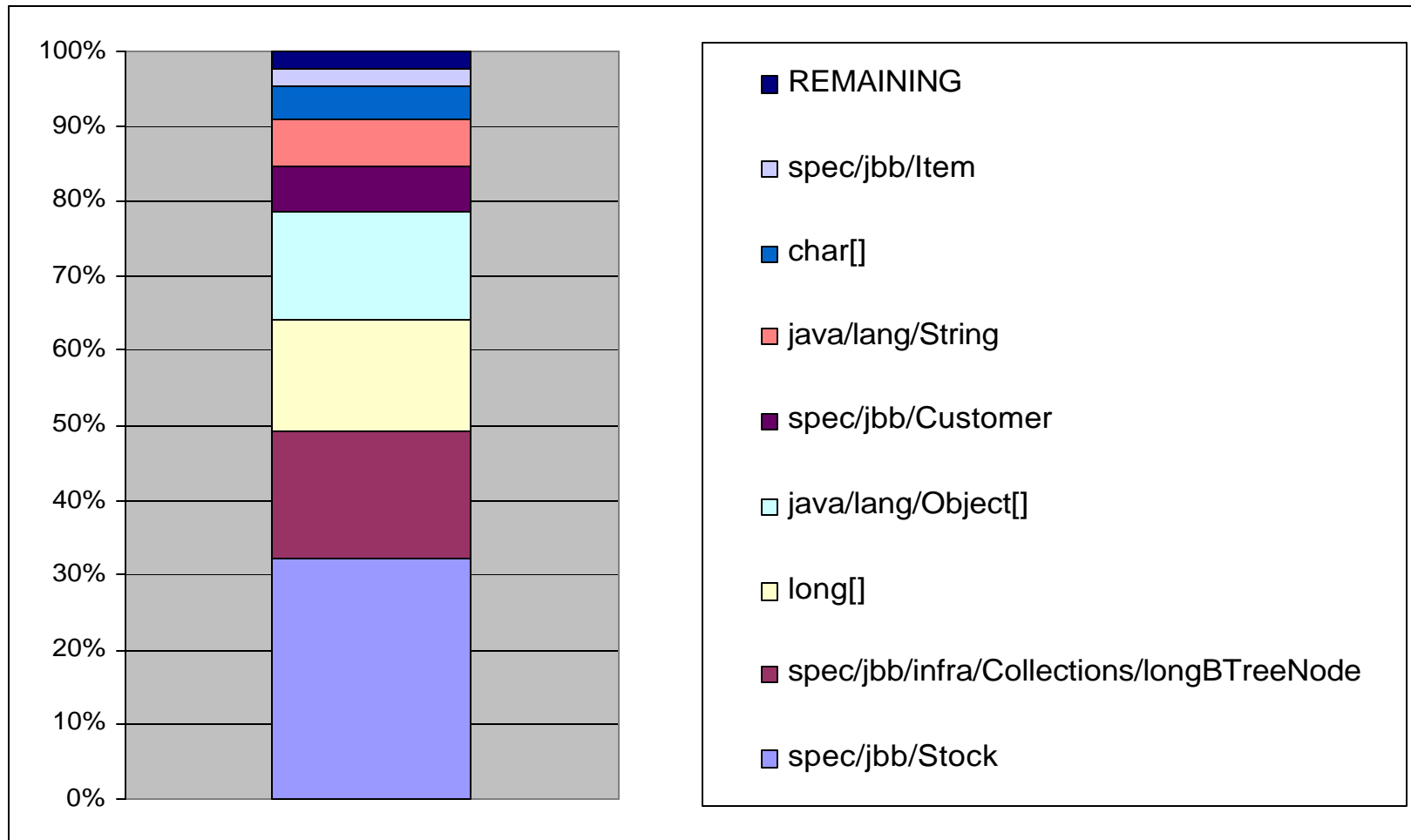
Instruction supply stall breakdown



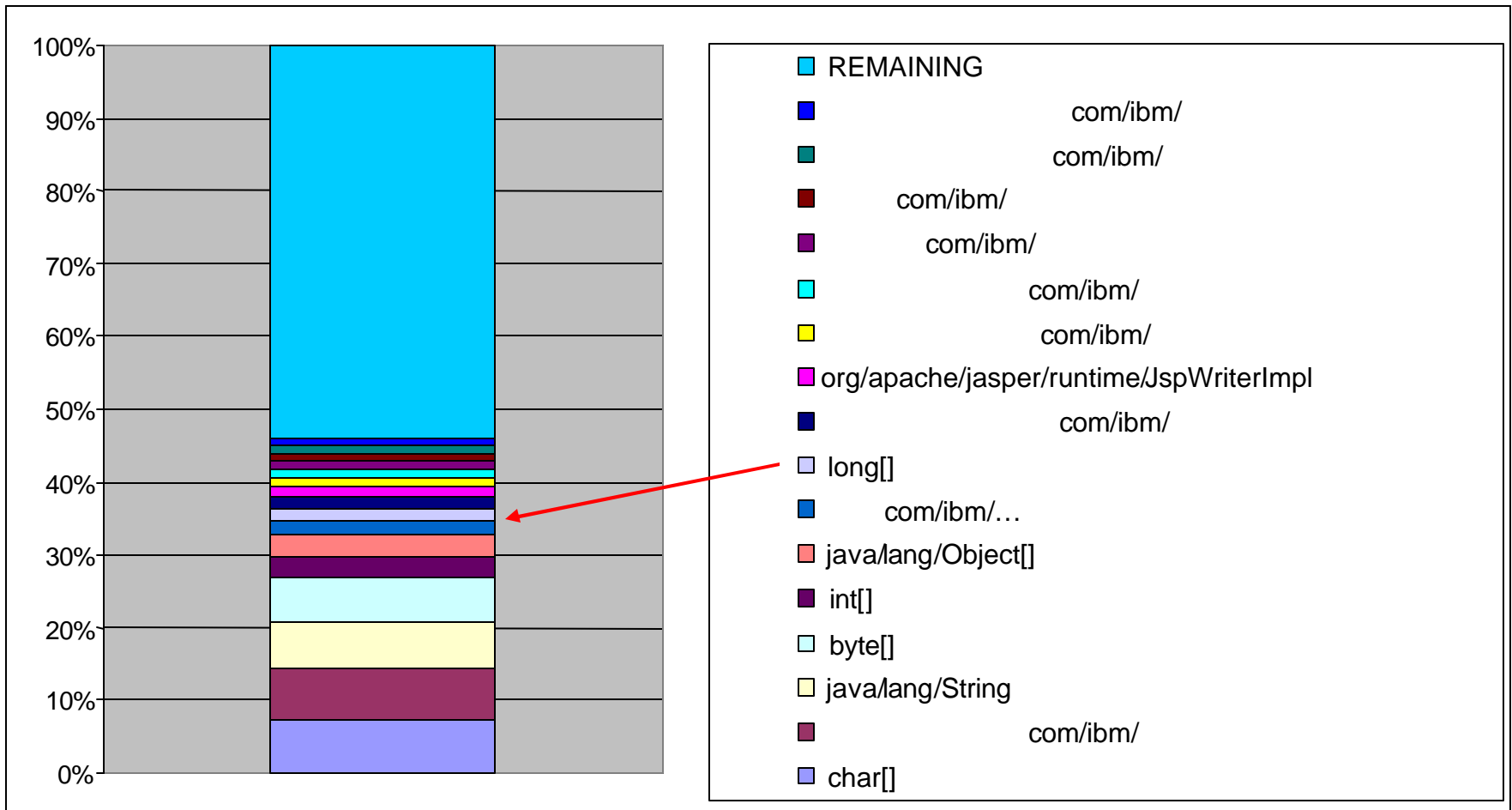
L1 Miss Data Load Patterns: JAS2004



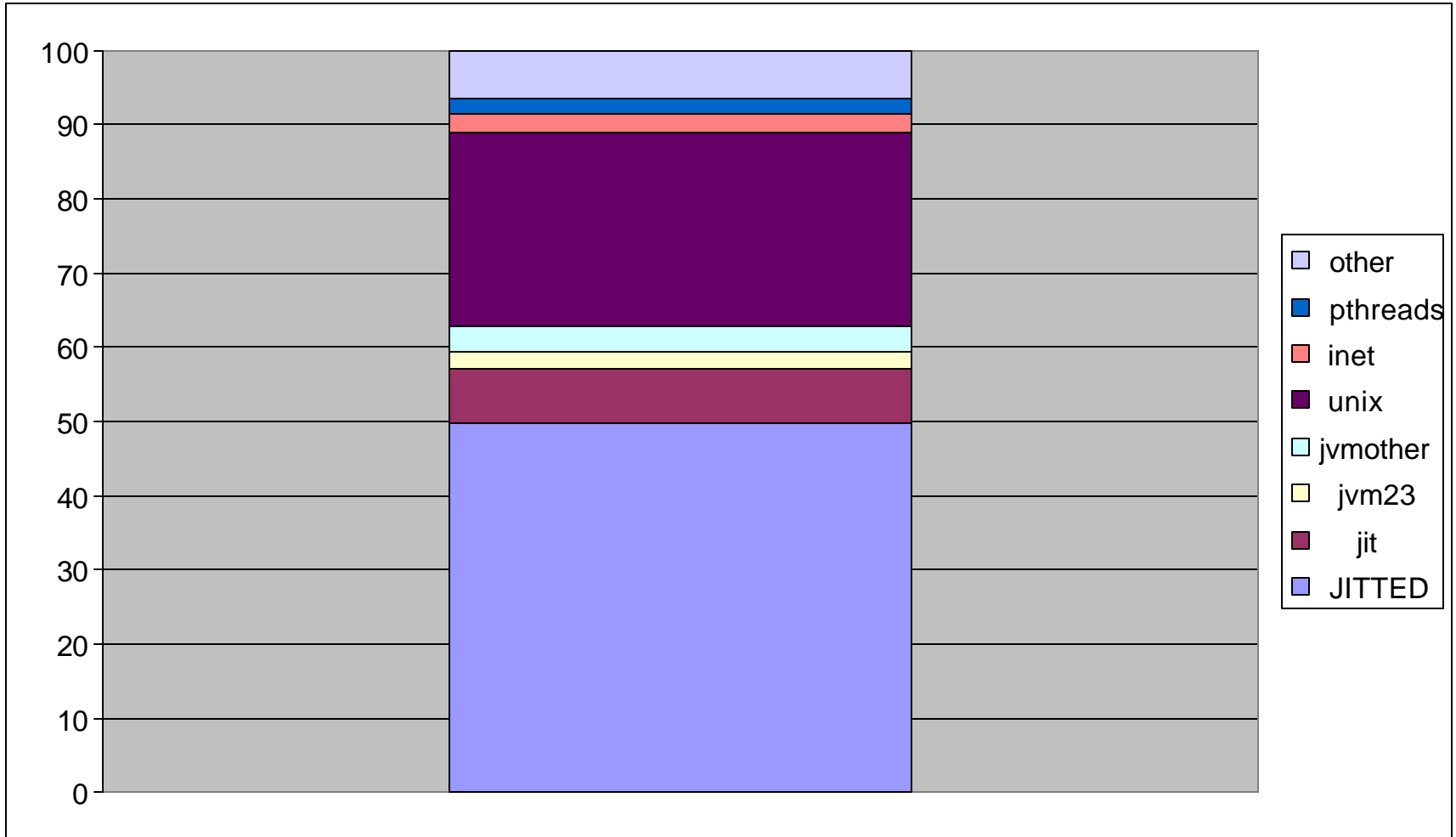
Types of Java Heap misses – SPECjbb2000



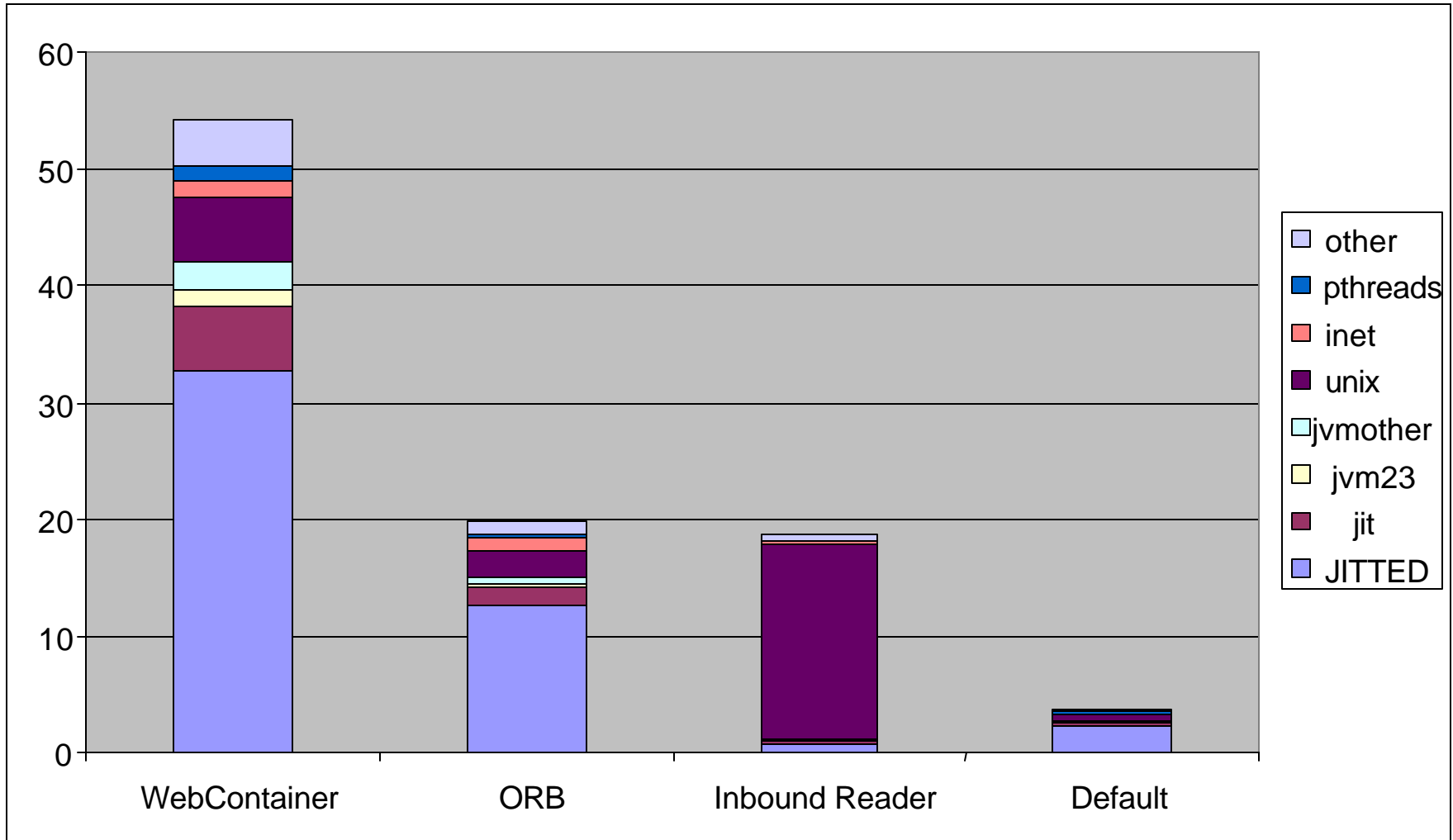
Types of Java Heap misses – JAS 2004



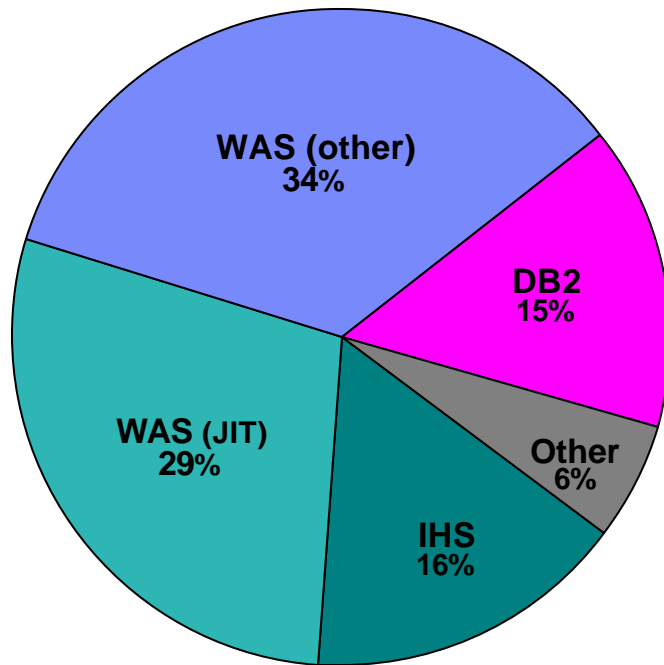
Misses by Component



Misses by Component

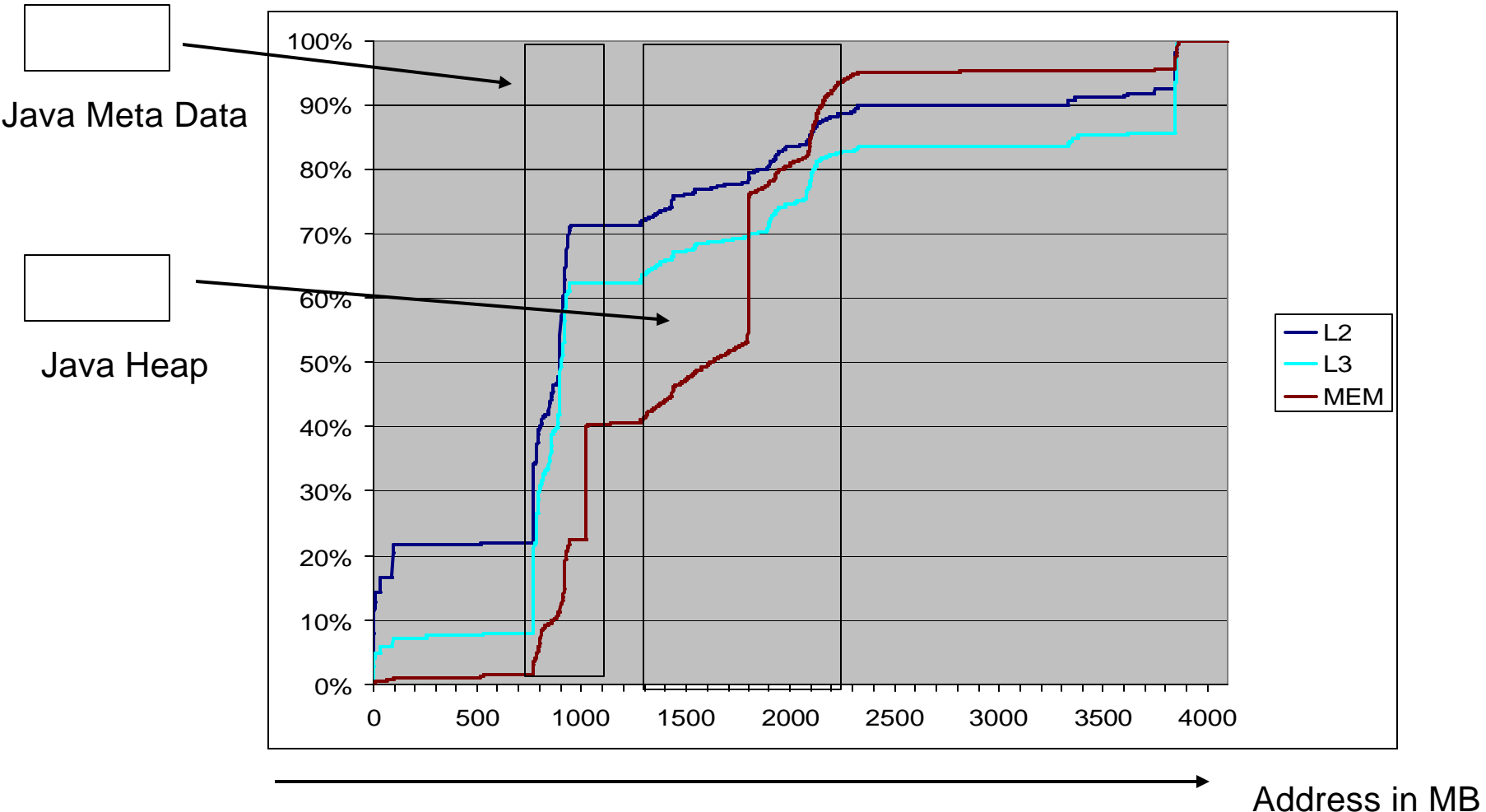


JIT Analysis

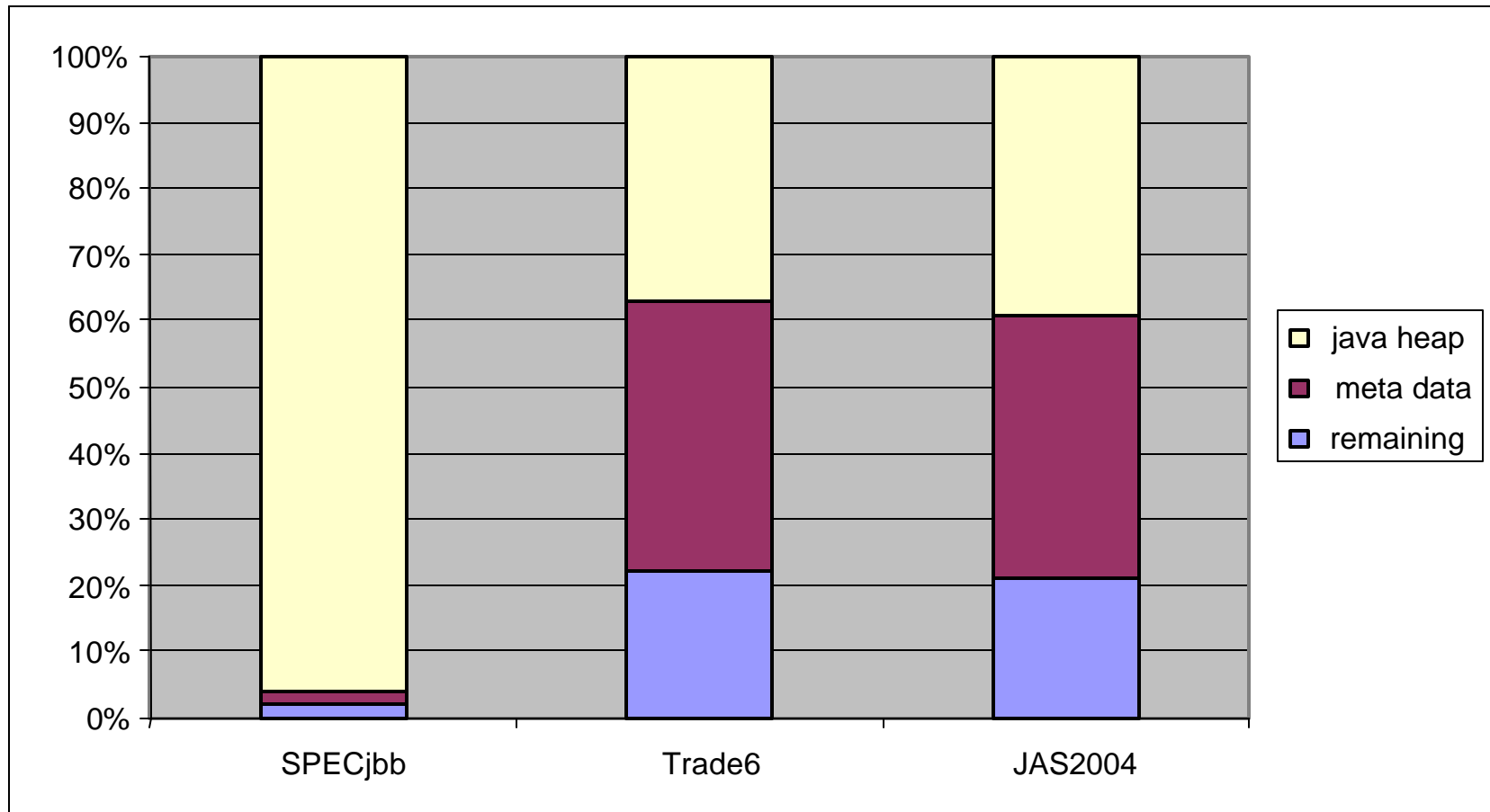


- **Data Collected from last 5 minutes of 60 minute run**
- **63% CPU time in WAS**
- **JIT'd Code in WAS (48% of WAS execution, 29% overall)**
 - Jas2004 JIT'd code: 3% of all JIT'd code
 - Enterprise Java Service <com.ibm.ejs>: 22% of JIT'd
 - WebSphere <com.ibm.ws>: 28% of JIT'd
- **Not-JIT'd Code in WAS (the other 52% of WAS execution time)**
 - 15% in kernel
 - 12% in libdb2.a
 - 11% in libmqmcs_r.a
 - 9% in libj9vm22.so
- **50% of JIT'd code execution is in 224 "hottest" methods**
 - Method profile is "flat"
 - Data profile is also "flat"

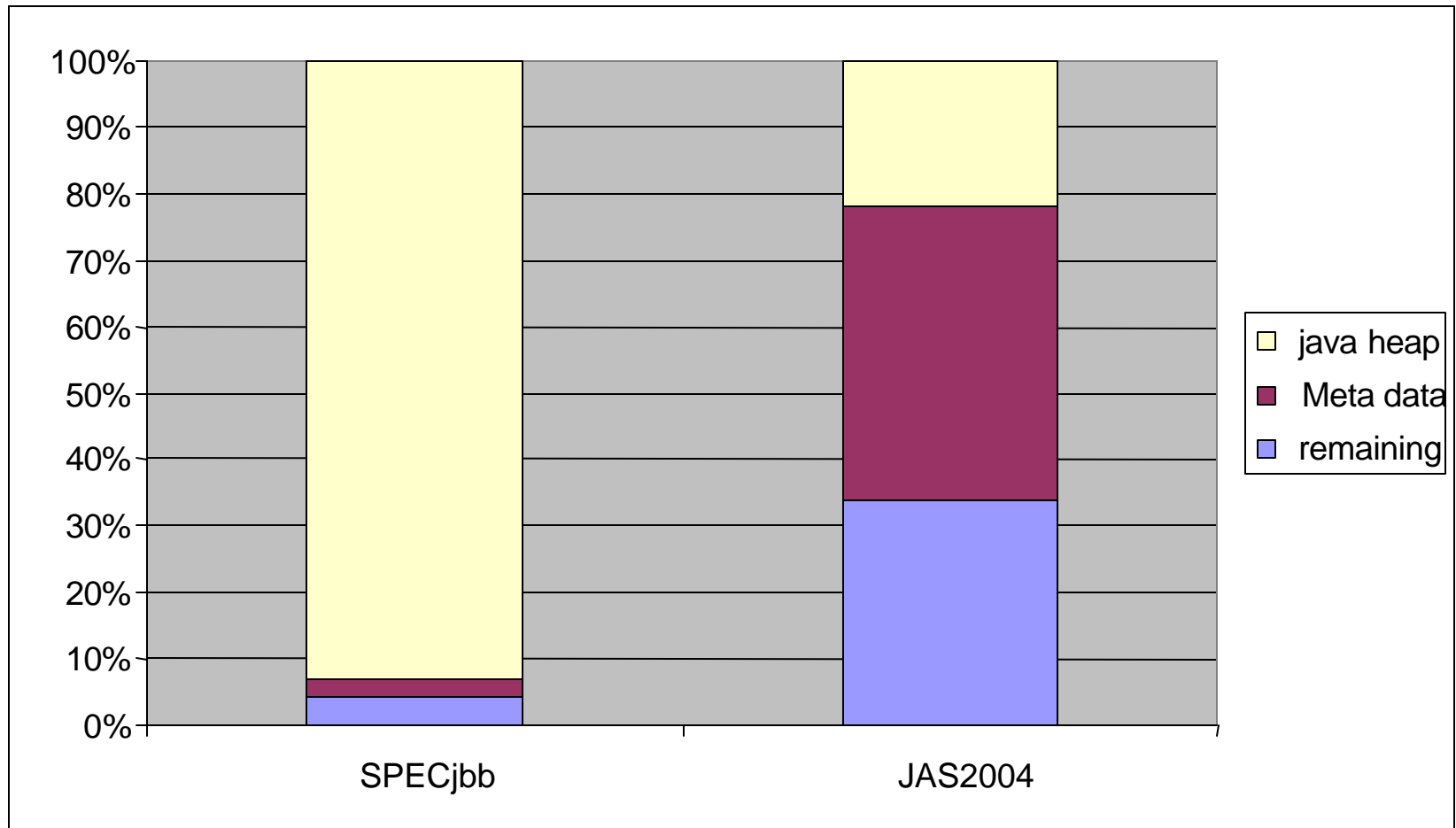
L1 Miss Data Load Patterns: JAS2004



Data Cache Misses



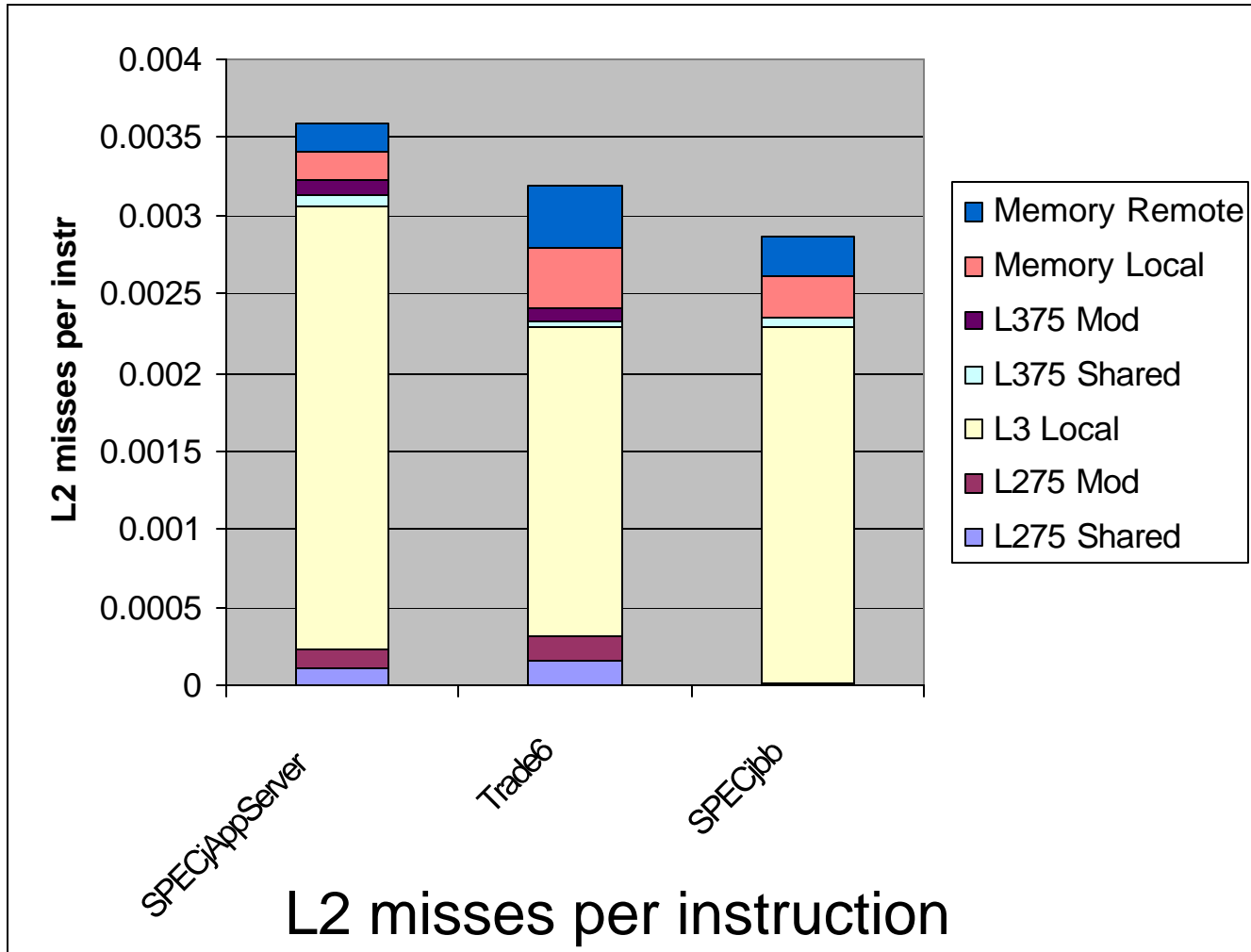
Loads from L3 Classified by Region



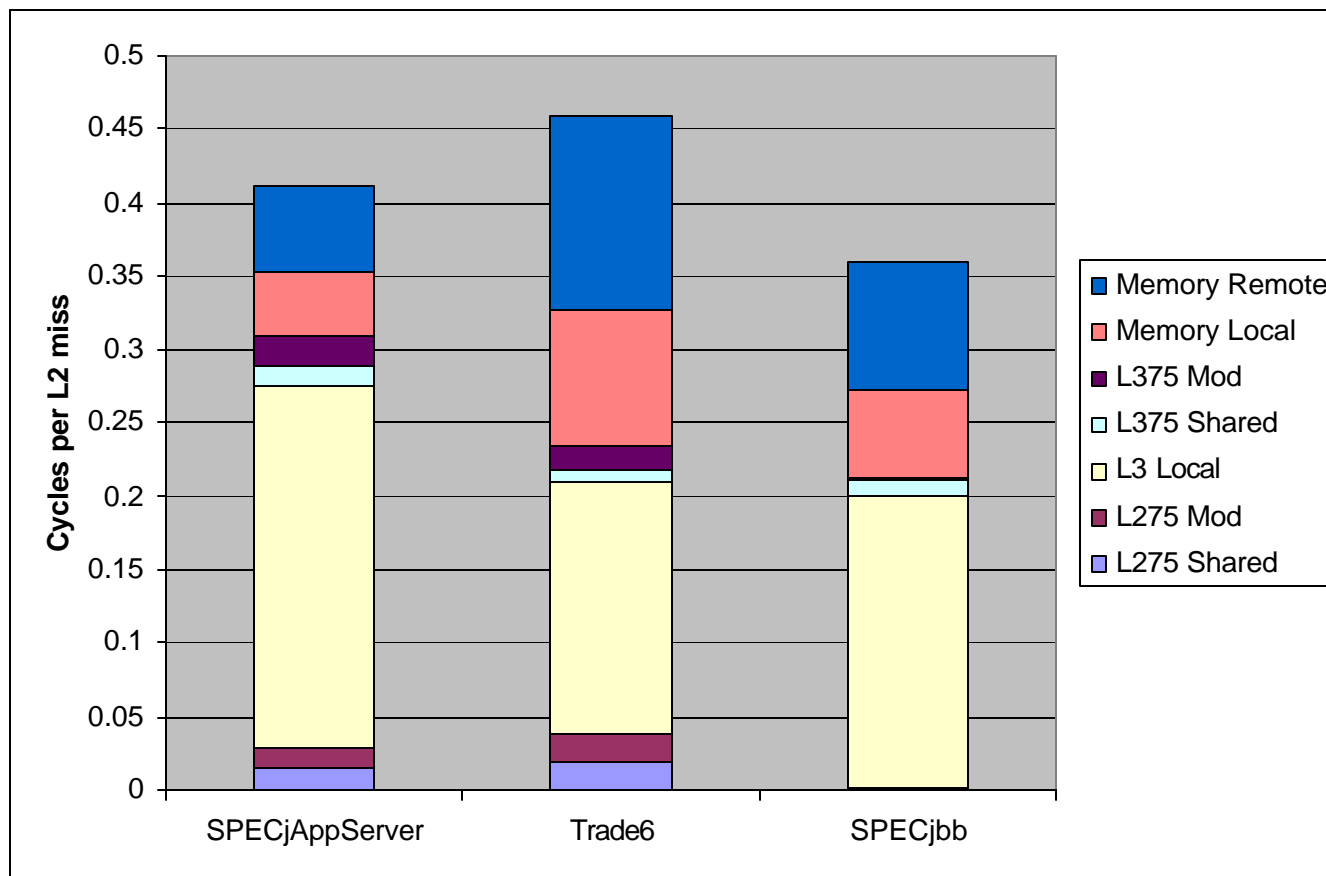
What is Meta-Data?

- JVM data structure not directly accessible by user application:
 - Object type information, Class information, Dispatch table, ...
 - Mostly accessed via indirection
 - Heavily used in Java
 - Invokevirtual, invokeinterface, checkcast, instanceof, ...

Capacity or Communication?



Capacity or Communication?

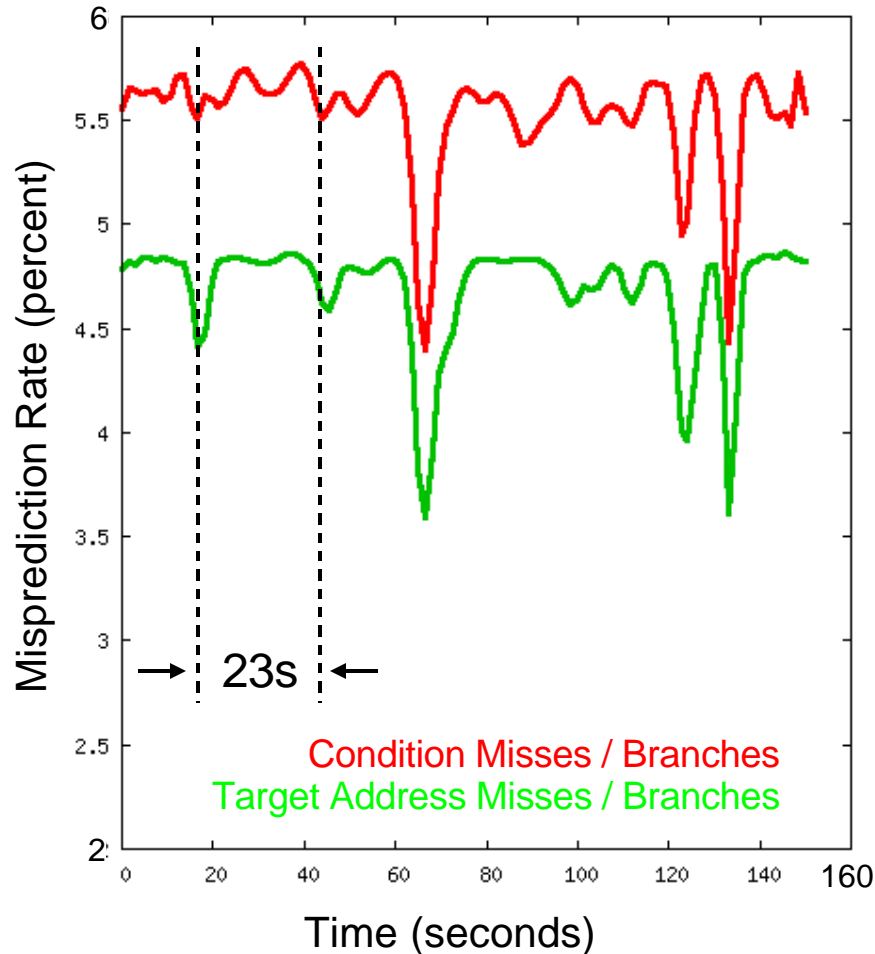


L2 miss cycles per instruction (approx.)

Data Cache Performance: Summary

- Memory ops are common - almost 50% of instructions
- Stronger Load performance, relatively weaker Store performance
- Mostly capacity misses w/o many communication misses
- Object Meta-Data accounts for a large portion of D\$ misses
 - Invokevirtual, invokeinterface, checkcast, instanceof, ...

Branch Misprediction



- **Relatively high as expected**
 - Correlated with GC events
- **Target Address (TA) Misses are strongly correlated with L1 I\$ miss rate (0.9)**
 - TA misses *could* lead to fetching useless instructions, evicting useful data & instructions
- **No apparent L1 D\$ pollution**
 - Low correlation between “speculation” rate and L1 D\$ misses

Relatively high mispr. rate, insignificant correlation with CPI

Comparison (Aggregate View), AIX/Power4

A: TrWasSovS+U: Trade3, WAS, Sovereign, System+User, 4 CPUs
B: TrWasJ9S+U: Trade3, WAS, J9, System+User, 4 CPUs
C: TrWasJ9U: Trade3, WAS, J9, User, 4 CPUs
D: JbbJ9S+U: SpecJBB, J9, System+User, 4 CPUs
E: TPC-C: Native (C) Code, 32 CPUs

	A: TrWasSovS+U	B: TrWasJ9S+U	C: TrWasJ9U	D: JbbJ9S+U	E: TPC-C
1: CPI	3.846	3.780	3.313	1.79	3.682
2: BR/Inst	24.43 %	23.38 %	23.27 %	18.96 %	19.0140 %
3: MPRED_CR/BR	6.48 %	5.23 %	5.28 %	5.36 %	4.3095 %
4: MPRED_TA/BR	4.96 %	4.64 %	5.38 %	1.35 %	1.7619 %
5: MPRED/BR	11.44 %	9.87 %	10.64 %	6.71 %	6.0714 %
6: MPRED/Inst:	2.79 %	2.27 %	2.48 %	1.27 %	1.1546 %
7: MPRED_CR/Inst	1.58 %	1.22 %	1.23 %	1.02 %	0.7619 %
8: MPRED_TA/Inst	1.21 %	1.09 %	1.25 %	0.26 %	0.3927 %

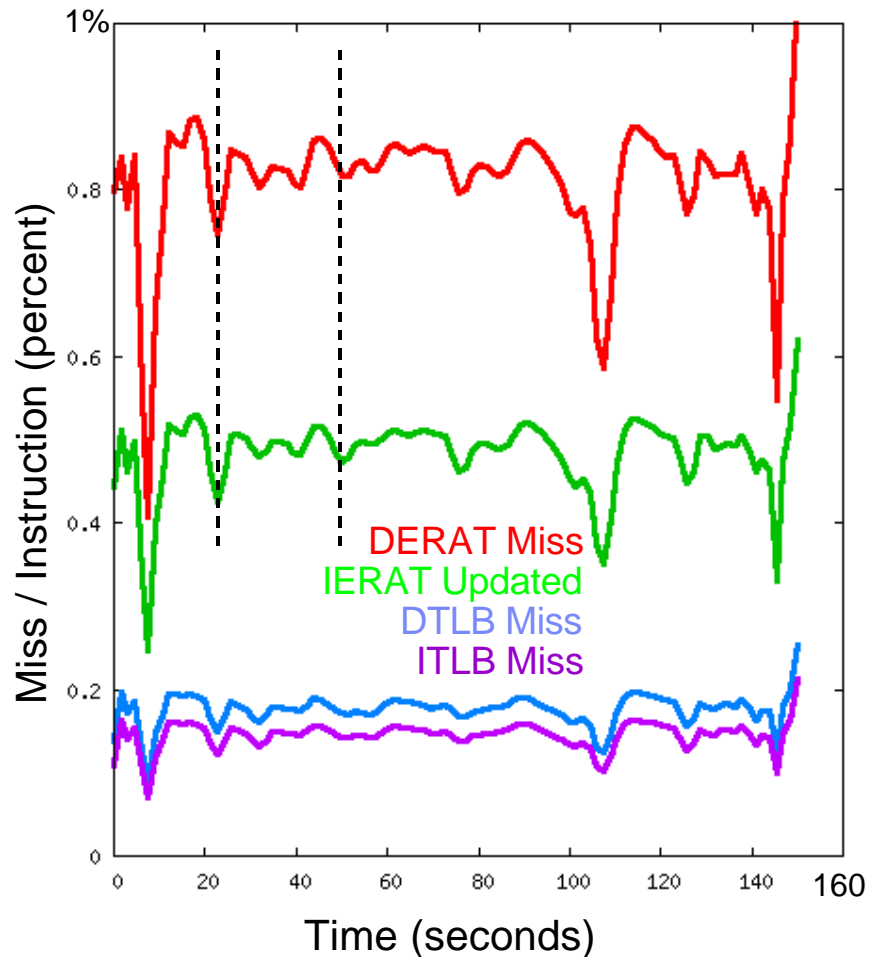
Small Java on J9 (D) shows very good CPI (1.79)

Branch Rate: WAS/apps (A – C) > small Java (D), Native code (E)

Branch Misprediction (CR: conditional): WAS/apps (A – C) > small Java (D), Native code (E), 2:1

Branch Misprediction (TA: target addr): WAS/apps (A – C) >> small Java (D), 4:1; Native (E), 3:1

Address Translation



Tolerable frequency of TLB & ERAT misses

- 2 - 3 orders of magnitude fewer TLB misses during GC
 - Graph fitted using Bezier smoothing -- spikes actually correspond to events that take 0.2 - 0.3s. -- the time of a GC
- ~500 instructions / DTLB miss
- ~25% of DERAT misses result in a TLB Miss → can be expensive

Large Pages help!

- DTLB Miss Rate improved by 25%
- ITLB Miss Rate improved by 15%

Much of working set is maintained in ERAT and TLB

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Synchronization Overhead

- Mem-sync instructions can occur **frequently** in multi-threaded server code.
- **Cost** is relatively high

	Power4	Power5
isync	30	10
lwsync	110	25
sync	140	50
lwarx/stwcx	80	75

Cycle times for memory barriers and atomic read/write

Locking Frequency

<i>benchmark</i>	<i>freq [ops/ms]</i>
jvm98_mtrt	120
jgf_mol	<< 1
jgf_monte	1650
jgf_ray	<<1
hedc	30
jigsaw	45
jbb (4wh)	1530
jbb (16 wh)	2320
trade6	630

Overhead in single-threaded benchmarks

- IBM's commercial VM on 4-way Power5 1.6 GHz
- Removed sync-operations in the JIT code generator

	<i>speedup</i>
jvm98_db	1.15
jvm98_jack	1.04
jvm98_javac	1.09
jgf_monte_A	1.02
jgf_monte_B	1.04
jbb (1 wh)	1.04

Thread-Local Locks

Fraction of lock operations on **thread-local** locks [%]:

jvm98_mtrt	99.3
jgf_mol	82.7
jgf_monte	99.6
jgf_ray	81.3
hedc	89.6
jigsaw	45.5
jbb (4wh)	33.5
jbb (16 wh)	18.9
trade6	30.3

Lock locality

scheduling dependence

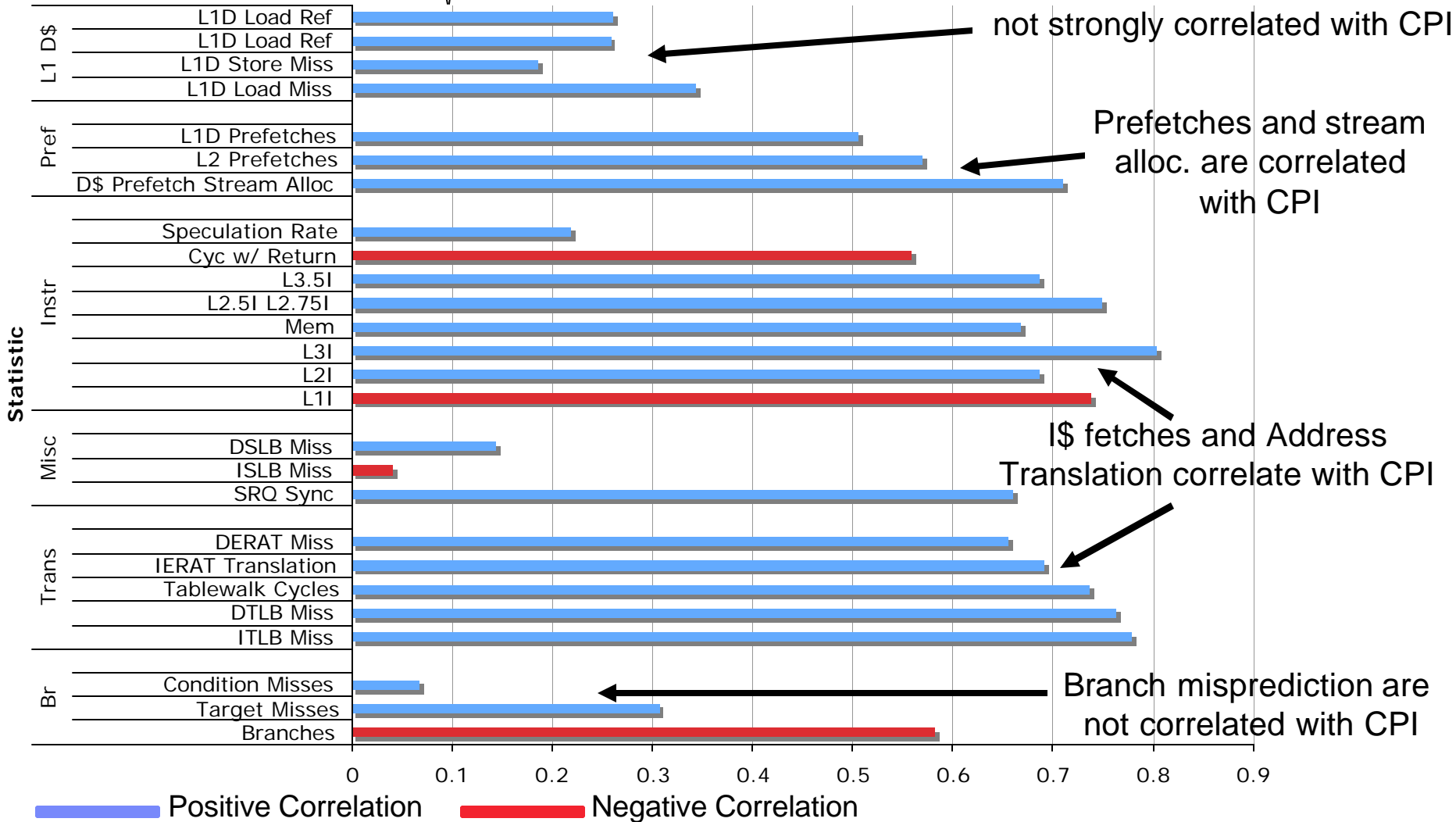
Dynamic fraction of threads that have locality to <...>:

implies thread-local

	<i>thr [%]</i>	<i>proc [%]</i>	<i>thr or proc [%]</i>
jvm98_mtrt	99.9	99.9	99.9
jgf_mol	98.4	98.2	98.5
jgf_monte	99.8	98.8	99.8
jgf_ray	99.0	99.0	99.2
hedc	97.2	97.6	98.1
jigsaw	84.8	91.8	91.8
jbb (4wh)	99.7	94.9	99.7
jbb (16 wh)	99.7	91.7	99.7
trade6	74.5	64.5	76.7

CPI Correlation

$$\frac{\sum(x-\bar{x})(y-\bar{y})}{\sqrt{\sum(x-\bar{x})^2 \sum(y-\bar{y})^2}}$$



L1 D\$ events not strongly correlated with CPI

Prefetches and stream alloc. are correlated with CPI

I\$ fetches and Address Translation correlate with CPI

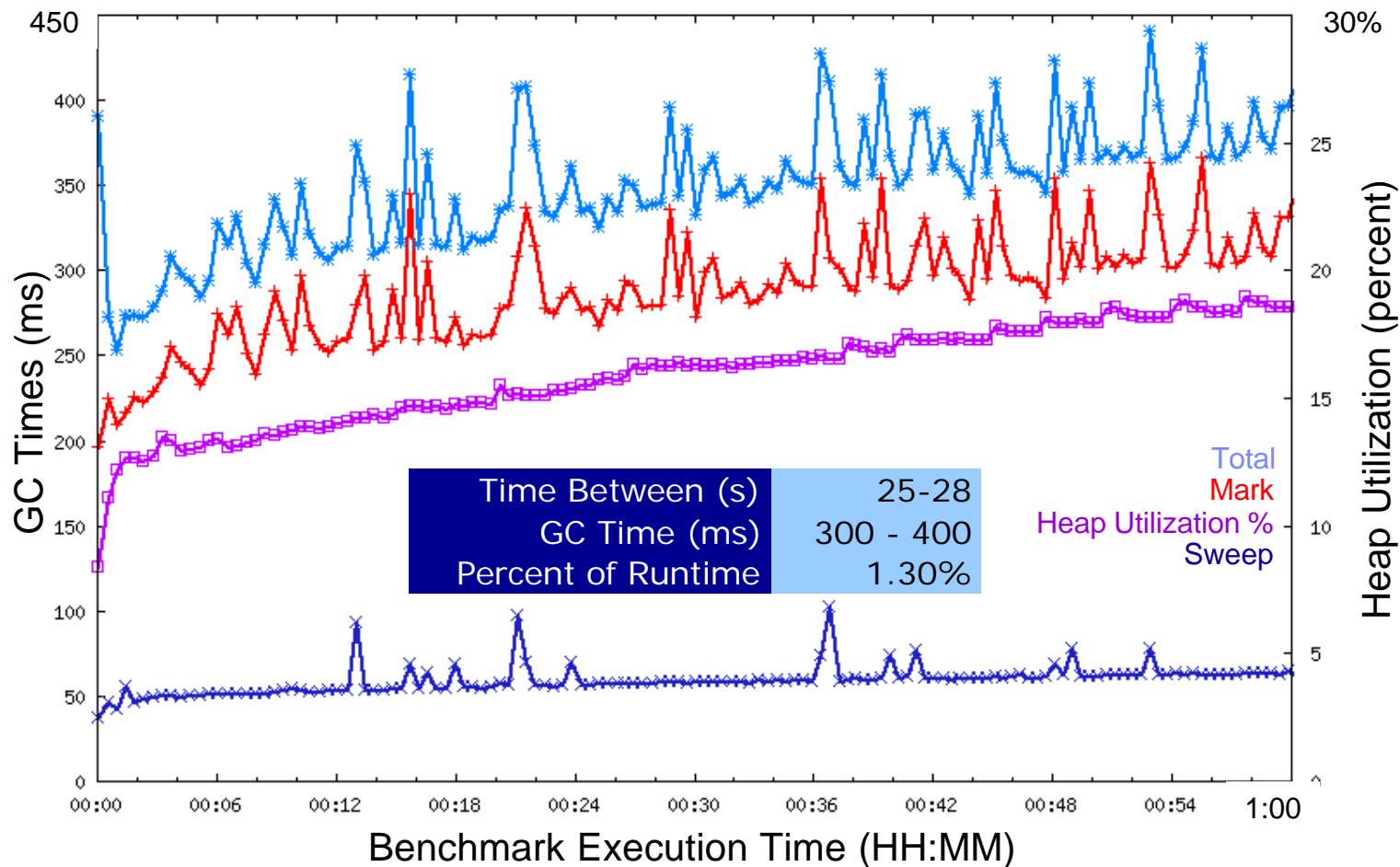
Branch misprediction are not correlated with CPI

No single parameter is perfectly correlated with CPI – a balanced system

Summary

- We have presented performance characteristics of Java server workloads
- Unlike a desktop system, GC is not a big issue
- They have higher branch-target misprediction rate
- Data cache miss rates are high
 - Java meta data cache miss is high
 - Mostly capacity misses, and low communication misses
- About 60% of CPU time is in Java, and ½ of that is in the jited code
- Method profile is flat
- Quite a large number of redundant memory sync operations
- No single performance metric has a very high correlation with CPI.

Garbage Collection



GC not as significant as past characterization papers have shown