Caches in WCET Analysis

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Outline

1 Caches

2 Cache Analysis for Least-Recently-Used

3 Beyond Least-Recently-Used
   - Predictability Metrics
   - Relative Competitiveness
   - Sensitivity – Caches and Measurement-Based Timing Analysis

4 Summary
Outline

1 Caches

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3 Beyond Least-Recently-Used
   - Predictability Metrics
   - Relative Competitiveness
   - Sensitivity – Caches and Measurement-Based Timing Analysis

4 Summary
Caches

- How they work:
  - dynamically
  - managed by replacement policy

![Diagram of CPU, Cache, and Main Memory]

- Capacity: 32 KB
- Latency: 3 cycles
- 2 MB, 100 cycles

- Why they work: *principle of locality*
  - spatial
  - temporal
Caches

How they work:
- dynamically
- managed by replacement policy

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Latency: 3 cycles

Why they work: principle of locality
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Caches

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```
CPU -> Cache [c3] -> Main Memory
```

- Capacity: 32 KB
- Latency: 3 cycles
- Capacity: 2 MB
- Latency: 100 cycles

- Why they work: *principle of locality*
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Caches

- How they work:
  - dynamically
  - managed by replacement policy

![Diagram of CPU, Cache, and Main Memory with "miss" and latency/capacity values]

- Why they work: *principle of locality*
  - spatial
  - temporal

Capacity:
- CPU: 32 KB
- Main Memory: 2 MB

Latency:
- CPU: 3 cycles
- Main Memory: 100 cycles
Caches

- How they work:
  - dynamically
  - managed by replacement policy

- Why they work: *principle of locality*
  - spatial
  - temporal

Diagram:
- CPU
- Cache
- Main Memory
- "miss" \[ ac \]
- Capacity: 32 KB
- Latency: 3 cycles
- 2 MB
- 100 cycles

Equation:
\[ c = \langle c_1 c_2 c_3 c_4 \rangle ! \]
Caches

- How they work:
  - dynamically
  - managed by replacement policy

```
Caches
CPU
Cache
Main Memory
```

```
Capacity: 32 KB
Latency: 3 cycles
```

```
Capacity: 2 MB
Latency: 100 cycles
```

```
“miss” [ac]
c3!
```

- Why they work: principle of locality
  - spatial
  - temporal
Caches

- How they work:
  - dynamically
  - managed by replacement policy

  ![Diagram of CPU, Cache, Main Memory with "hit" and latency/capacity values]

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  - temporal
Caches

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![Diagram of CPU, Cache, and Main Memory with capacities and latencies]

- Why they work: *principle of locality*
  - spatial
  - temporal
Fully-Associative Caches

Address:

<table>
<thead>
<tr>
<th>Tag</th>
<th>Block offset</th>
</tr>
</thead>
</table>

$B_b k s \log_2 (s) \log_2 (8 \times b)$

MUX Data

$= \text{associativity}$

Tag Data Block

Yes: Hit!

No: Miss!

$k = \text{associativity}$

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Set-Associative Caches

Special cases:
- **direct-mapped cache**: only one line per cache set
- **fully-associative cache**: only one cache set
Cache Replacement Policies

- Least-Recently-Used (LRU) used in
  Intel Pentium I and MIPS 24K/34K

- First-In First-Out (FIFO or Round-Robin) used in
  Motorola POWERPC 56X, Intel XScale, ARM9, ARM11

- Pseudo-LRU (PLRU) used in
  Intel Pentium II-IV and POWERPC 75X

- Most Recently Used (MRU) as described in literature

Each cache set is treated independently:

→ Set-associative caches are compositions of fully-associative caches.
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Cache Analysis

Two types of cache analyses:

1. Local guarantees: classification of individual accesses
   - May-Analysis \(\rightarrow\) Overapproximates cache contents
   - Must-Analysis \(\rightarrow\) Underapproximates cache contents

2. Global guarantees: bounds on cache hits/misses

- Cache analyses almost exclusively for LRU
- In practice: FIFO, PLRU, \ldots
Challenges for Cache Analysis

Always a cache hit/always a miss?

1. Initial cache contents unknown.
2. Different paths lead to these points.
3. Cannot resolve address of $z$. 

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Challenges for Cache Analysis

- Always a cache hit/always a miss?
  1. Initial cache contents unknown.
  2. Different paths lead to these points.
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Deriving Invariants about Cache States using Abstract Interpretation

**Collecting Semantics** = set of states at each program point that any execution may encounter there

Two approximations:

- Collecting Semantics \subseteq \text{uncomputable}
- Cache Semantics \subseteq \text{computable}
- γ(Abstract Cache Sem.) \subseteq \text{efficiently computable}
Deriving Invariants about Cache States using Abstract Interpretation

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Two approximations:
- Collecting Semantics uncomputable
- ⊆ Cache Semantics computable
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Collecting Semantics = set of states at each program point that any execution may encounter there

Two approximations:
- Collecting Semantics \( \subseteq \) Cache Semantics, computable
- Collecting Semantics \( \subseteq \gamma(\text{Abstract Cache Sem.}) \), efficiently computable
Deriving Invariants about Cache States using Abstract Interpretation

Collecting Semantics = set of states at each program point that any execution may encounter there

Two approximations:
- Collecting Semantics uncomputable
- ⊆ Cache Semantics computable
- ⊆ γ(Abstract Cache Sem.) efficiently computable
Least-Recently-Used (LRU): Concrete Behavior

“Cache Miss”:

```
  z y x t
  s    
```

LRU has notion of age

“Cache Hit”:

```
  s z y t
  s    
```

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LRU: Must-Analysis: Abstract Domain

- Used to predict *cache hits*.
- Maintains *upper bounds on ages* of memory blocks.
- Upper bound $\leq$ associativity $\rightarrow$ memory block definitely cached.

**Example**

<table>
<thead>
<tr>
<th>Abstract state:</th>
<th>... and its interpretation:</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Describes the set of all concrete cache states in which $x$, $s$, and $t$ occur,</td>
</tr>
<tr>
<td></td>
<td>- $x$ with an age of 0,</td>
</tr>
<tr>
<td></td>
<td>- $s$ and $t$ with an age not older than 2.</td>
</tr>
<tr>
<td>{x}</td>
<td>$\gamma([{x}, {}, {s, t}, {}]) =$</td>
</tr>
<tr>
<td>{}</td>
<td>${[x, s, t, a], [x, t, s, a], [x, s, t, b], \ldots}$</td>
</tr>
</tbody>
</table>
Sound Update – Local Consistency

Abstract Update

\((\text{must})\) \quad \rightarrow \quad (\text{must}')\)

\(\gamma\)

concrete cache states

Lifted Concrete Update

concrete cache states

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LRU: Must-Analysis: Update

“Potential Cache Miss”:

“Definite Cache Hit”:

Why does $t$ not age in the second case?
LRU: Must-Analysis: Join

Need to combine information where control-flow merges.

Join should be conservative:

- $\gamma(A) \subseteq \gamma(A \sqcup B)$
- $\gamma(B) \subseteq \gamma(A \sqcup B)$

```
\begin{array}{|c|c|c|}
\hline
\{a\} & \{c\} & \{\}\ \\
\{\} & \{e\} & \{\}\ \\
\{c,f\} & \{a\} & \{a,c\} \\
\{d\} & \{d\} & \{d\} \\
\hline
\end{array}
```

“Intersection + Maximal Age”
LRU: Must-Analysis: Join

Need to combine information where control-flow merges.

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- $\gamma(A) \subseteq \gamma(A \sqcup B)$
- $\gamma(B) \subseteq \gamma(A \sqcup B)$

```
\begin{array}{c|c|c}
<p>| | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>a</td>
<td>c</td>
<td></td>
</tr>
<tr>
<td></td>
<td>e</td>
<td>a</td>
</tr>
<tr>
<td></td>
<td></td>
<td>d</td>
</tr>
<tr>
<td></td>
<td>c</td>
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</tbody>
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\end{array}
```

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Join should be conservative:

- \( \gamma(A) \subseteq \gamma(A \cup B) \)
- \( \gamma(B) \subseteq \gamma(A \cup B) \)

```
  \{a\}   \{c\}   \{\} \\
  \{\}   \{e\}   \{\} \\
 \{c,f\} \{a\}   \{a\} \\
 \{d\}   \{d\}   \{d\}
```

“Intersection + Maximal Age”
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Need to combine information where control-flow merges.

Join should be conservative:

- \( \gamma(A) \subseteq \gamma(A \cup B) \)
- \( \gamma(B) \subseteq \gamma(A \cup B) \)

```
| {a}   | {c}   | {}    |
| {e}   | {a}   | {}    |
| {c,f} | {a,c} | {d}   |
| {d}   | {d}   | {d}   |
```

“Intersection + Maximal Age”
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Need to combine information where control-flow merges.

Join should be conservative:

- $\gamma(A) \subseteq \gamma(A \sqcup B)$
- $\gamma(B) \subseteq \gamma(A \sqcup B)$

```
\begin{array}{c}
\{a\} \\
\{\} \\
\{c,f\} \\
\{d\}
\end{array}
\sqcup
\begin{array}{c}
\{c\} \\
\{e\} \\
\{a\} \\
\{d\}
\end{array}
\sqcup
\begin{array}{c}
\{\} \\
\{\} \\
\{a,c\} \\
\{d\}
\end{array}
```

“Intersection + Maximal Age”
LRU: Must-Analysis: Join

Need to combine information where control-flow merges.

Join should be conservative:

- $\gamma(A) \subseteq \gamma(A \sqcup B)$
- $\gamma(B) \subseteq \gamma(A \sqcup B)$

```
{a}    {c}    {}
{e}    {e}    {}
{c.f}  {a}    {a,c}
{d}    {d}    {d}
```

“Intersection + Maximal Age”
LRU: Must-Analysis: Join

Need to combine information where control-flow merges.

Join should be conservative:
- \( \gamma(A) \subseteq \gamma(A \sqcup B) \)
- \( \gamma(B) \subseteq \gamma(A \sqcup B) \)

```
\{a\}  \sqcup \{c\}  \sqcup \{\}  \sqcup \{\}\n\{\}  \sqcup \{e\}  \sqcup \{\}\n\{c,f\} \sqcup \{a\}  \sqcup \{a,c\}\n\{d\}  \sqcup \{d\}  \sqcup \{d\}
```

“Intersection + Maximal Age”

How many memory blocks can be in the must-cache?
Example: Must-Analysis

\[
\text{entry } \{\}, \{\}, \{\}, \{\}
\]

\[
\text{exit } \bot
\]

\[
A \bot B \bot C \bot D
\]
Example: Must-Analysis

\[
\begin{align*}
\text{entry } [\{\}, \{\}, \{\}, \{\}] \\
\downarrow \sqcup [\{\}, \{\}, \{\}, \{\}] = [\{\}, \{\}, \{\}, \{\}]
\end{align*}
\]
Example: Must-Analysis

\[
\text{entry } \{\}, \{\}, \{\}, \{\}, \{\} = \{\}, \{\}, \{\}, \{\}
\]

\[
\{\}, \{\}, \{\}, \{\}, \{\}
\]

\[
\{\}, \{\}, \{\}, \{\}, \{\}
\]

\[
\{\}, \{\}, \{\}, \{\}, \{\}
\]

\[
\{\}, \{\}, \{\}, \{\}, \{\}
\]

\[
\{\}, \{\}, \{\}, \{\}, \{\}
\]

\[
\{\}, \{\}, \{\}, \{\}, \{\}
\]

\[
\{\}, \{\}, \{\}, \{\}, \{\}
\]
Example: Must-Analysis

\[ \{\}, \{\}, \{\}, \{\} \]

A \perp \sqcup \{\}, \{\}, \{\}, \{\} = \{\}, \{\}, \{\}, \{\}

\[ \{A\}, \{\}, \{\}, \{\} \]

B \rightarrow A

\[ \{B\}, \{A\}, \{\}, \{\} \sqcup \{C\}, \{A\}, \{\}, \{\} = \{\}, \{A\}, \{\}, \{\} \]

D \rightarrow C

\[ \{A\}, \{\}, \{\}, \{\} \]

C \rightarrow D

exit \perp
Example: Must-Analysis

entry $[\{\},\{\},\{\},\{\}]$

$[\{D\},\{\},\{A\},\{\}] \sqcup [\{\},\{\},\{\},\{\}] = [\{\},\{\},\{\},\{\}]$

$[\{A\},\{\},\{\},\{\}]

$[\{B\},\{A\},\{\},\{\}] \sqcup [\{C\},\{A\},\{\},\{\}] = [\{\},\{A\},\{\},\{\}]

exit $[\{D\},\{\},\{A\},\{\}]$

No cache hits can be predicted :-(
Context-Sensitive Analysis/Virtual Loop-Unrolling

- **Problem:**
  - The first iteration of a loop will always result in cache misses.
  - Similarly for the first execution of a function.

- **Solution:**
  - Virtually Unroll Loops: Distinguish the first iteration from others
  - Distinguish function calls by calling context.

Virtually unrolling the loop once:

- **Accesses to** $A$ **and** $D$ **are provably hits after the first iteration**

- **Accesses to** $B$ **and** $C$ **can still not be classified. Within each execution of the loop, they may only miss once.**

  → Persistence Analysis
LRU: May-Analysis: Abstract Domain

- Used to predict *cache misses*.
- Maintains *lower bounds on ages* of memory blocks.
- Lower bound $\geq$ associativity

$$\rightarrow$$ memory block definitely *not* cached.

### Example

<table>
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</thead>
<tbody>
<tr>
<td>${x, y}$</td>
<td>Describes the set of all concrete cache states in which no memory blocks except $x$, $y$, $s$, $t$, and $u$ occur,</td>
</tr>
<tr>
<td>age 0</td>
<td>- $x$ and $y$ with an age of at least 0,</td>
</tr>
<tr>
<td>{}</td>
<td>- $s$ and $t$ with an age of at least 2,</td>
</tr>
<tr>
<td>{s, t}</td>
<td>- $u$ with an age of at least 3.</td>
</tr>
<tr>
<td>{u}</td>
<td>$\gamma({{x, y}, {}, {s, t}, {u}}) = {[x, y, s, t], [y, x, s, t], [x, y, s, u], \ldots}$</td>
</tr>
</tbody>
</table>
LRU: May-Analysis: Update

“Definite Cache Miss”:

```
\{x\}  \{z\}
\{\}   \{x\}
\{s,t\}\{\}
 \{y\}  \{s,t\}
```

“Potential Cache Hit”:

```
\{x\}  \{s\}
\{\}   \{x\}
\{s,t\}\{\}
 \{y\}  \{y,t\}
```

Why does \( t \) age in the second case?
LRU: May-Analysis: Join

Need to combine information where control-flow merges.

Join should be conservative:

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```
\begin{array}{c|c|c}
\{a\} & \{c\} & \{a,c\} \\
\{\} & \{e\} & \{e\} \\
\{c,f\} & \{a\} & \{f\} \\
\{d\} & \{d\} & \{d\}
\end{array}
```

“Union + Minimal Age”
Need to combine information where control-flow merges.

Join should be conservative:

- \( \gamma(A) \subseteq \gamma(A \cup B) \)
- \( \gamma(B) \subseteq \gamma(A \cup B) \)

\[
\begin{array}{c}
\{a\} \\
\{\} \\
\{c,f\} \\
\{d\}
\end{array}
\sqcup
\begin{array}{c}
\{c\} \\
\{e\} \\
\{a\} \\
\{d\}
\end{array}
= \begin{array}{c}
\{a,c\} \\
\{e\} \\
\{f\} \\
\{d\}
\end{array}
\]

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```
\{a\} | \{c\} | \{a,c\
\{\}  | \{e\}  | \{e\}
\{c,f\} | \{a\}  | \{f\}
\{d\}  | \{d\}  | \{d\}
```

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```
\begin{array}{c}
{a} \\
{} \\
{c,f} \\
{d}
\end{array} \quad \sqcup \quad \begin{array}{c}
{c} \\
{e} \\
{a} \\
{d}
\end{array} = \begin{array}{c}
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\[
\begin{array}{c}
{a} \\
{e} \\
{a,c} \\
{e,f}
\end{array} \quad \sqcup \quad \begin{array}{c}
{c} \\
{e} \\
{a} \\
{d}
\end{array} = \begin{array}{c}
{a,c} \\
{e} \\
{f} \\
{d}
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\[
\begin{align*}
\{a\} & \quad \{c\} & \quad \{a,c\} \\
\{\} & \quad \{e\} & \quad \{e\} \\
\{c,f\} & \quad \{a\} & \quad \{f\} \\
\{d\} & \quad \{d\} & \quad \{d\}
\end{align*}
\]

"Union + Minimal Age"
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Uncertainty in WCET Analysis

- Amount of uncertainty determines precision of WCET analysis
- Uncertainty in cache analysis depends on replacement policy

\[
\text{uncertainty} \times \text{penalty} \times \text{variation due to inputs and initial hardware state} \leq \text{WCET upper bound} \leq \text{execution time}
\]
Uncertainty in Cache Analysis

1. Initial cache contents unknown.
2. Need to combine information.
3. Cannot resolve address of \( z \).

\[ \Rightarrow \]

Amount of uncertainty determined by ability to recover information.

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Uncertainty in Cache Analysis

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Uncertainty in Cache Analysis

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Uncertainty in Cache Analysis

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Uncertainty in Cache Analysis

1. Initial cache contents unknown.
2. Need to combine information.
3. Cannot resolve address of \( z \).

\[ \Rightarrow \text{Amount of uncertainty determined by ability to recover information} \]
Predictability Metrics

Evict

Fill

Sequence: \( \langle a, \ldots, e, f, g, h \rangle \)
Meaning of Metrics

■ Evict
  ▶ Number of accesses to obtain any may-information.
  ▶ I.e. when can an analysis predict any cache misses?

■ Fill
  ▶ Number of accesses to complete may- and must-information.
  ▶ I.e. when can an analysis predict each access?

→ Evict and Fill bound the precision of any static cache analysis. Can thus serve as a benchmark for analyses.
Evaluation of Least-Recently-Used

- LRU “forgets” about past quickly:
  - cares about most-recent access to each block only
  - order of previous accesses irrelevant

![Diagram showing cache evictions and fills](image)

- In the example: \( \text{Evict} = \text{Fill} = 4 \)
- In general: \( \text{Evict}(k) = \text{Fill}(k) = k \), where \( k \) is the associativity of the cache
Evaluation of First-In First-Out (sketch)

- Like LRU in the miss-case
- But: “Ignores” hits

In the worst-case $k - 1$ hits and $k$ misses: $(k = \text{associativity})$
\[ \rightarrow \text{Evict}(k) = 2k - 1 \]

Another $k$ accesses to obtain complete knowledge:
\[ \rightarrow \text{Fill}(k) = 3k - 1 \]
Evaluation of Pseudo-LRU (sketch)

- Tree-bits point to block to be replaced

\[
\begin{array}{cccc}
1 & 1 & c & 1 \\
1 & a & b & c & d \\
0 & 1 & e & 1 \\
0 & a & e & c & d \\
\end{array}
\]

- Accesses “rejuvenate” neighborhood
  - Active blocks keep their (inactive) neighborhood in the cache

- Analysis yields:
  - \( \text{Evict}(k) = \frac{k}{2} \log_2 k + 1 \)
  - \( \text{Fill}(k) = \frac{k}{2} \log_2 k + k - 1 \)


## Evaluation of Policies

<table>
<thead>
<tr>
<th>Policy</th>
<th>Evict(k)</th>
<th>Fill(k)</th>
<th>Evict(8)</th>
<th>Fill(8)</th>
</tr>
</thead>
<tbody>
<tr>
<td>LRU</td>
<td>(k)</td>
<td>(k)</td>
<td>8</td>
<td>8</td>
</tr>
<tr>
<td>FIFO</td>
<td>(2k - 1)</td>
<td>(3k - 1)</td>
<td>15</td>
<td>23</td>
</tr>
<tr>
<td>MRU</td>
<td>(2k - 2)</td>
<td>(\infty / 3k - 4)</td>
<td>14</td>
<td>(\infty / 20)</td>
</tr>
<tr>
<td>PLRU</td>
<td>(\frac{k}{2} \log_2 k + 1)</td>
<td>(\frac{k}{2} \log_2 k + k - 1)</td>
<td>13</td>
<td>19</td>
</tr>
</tbody>
</table>

- LRU is optimal w.r.t. metrics.
- Other policies are much less predictable.

→ Use LRU if predictability is a concern.

- How to obtain *may-* and *must-*information within the given limits for other policies?
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Relative Competitiveness

- **Competitiveness** (Sleator and Tarjan, 1985): worst-case performance of an online policy *relative to the optimal offline policy*
  - used to evaluate online policies

- **Relative competitiveness** (Reineke and Grund, 2008): worst-case performance of an online policy *relative to another online policy*
  - used to derive local and global cache analyses
Definition – Relative Miss-Competitiveness

**Notation**

\[ m_P(p, s) = \text{number of misses that policy } P \text{ incurs on access sequence } s \in M^* \text{ starting in state } p \in C_P \]
Definition – Relative Miss-Competitiveness

Notation

\[ m_P(p, s) = \text{number of misses that policy } P \text{ incurs on access sequence } s \in M^* \text{ starting in state } p \in C_P \]

Definition (Relative miss competitiveness)

Policy \( P \) is \((k, c)\)-miss-competitive relative to policy \( Q \) if

\[ m_P(p, s) \leq k \cdot m_Q(q, s) + c \]

for all access sequences \( s \in M^* \) and cache-set states \( p \in C_P, q \in C_Q \) that are compatible \( p \sim q \).
Definition – Relative Miss-Competitiveness

Notation

\[ m_P(p, s) = \text{number of misses that policy } P \text{ incurs on} \]
\[ \text{access sequence } s \in M^* \text{ starting in state } p \in C^P \]

Definition (Relative miss competitiveness)

Policy \( P \) is \((k, c)\)-miss-competitive relative to policy \( Q \) if

\[ m_P(p, s) \leq k \cdot m_Q(q, s) + c \]

for all access sequences \( s \in M^* \) and cache-set states \( p \in C^P, q \in C^Q \) that are compatible \( p \sim q \).

Definition (Competitive miss ratio of \( P \) relative to \( Q \))

The smallest \( k \), s.t. \( P \) is \((k, c)\)-miss-competitive rel. to \( Q \) for some \( c \).
Example – Relative Miss-Competitiveness

\( P \) is \((3, 4)\)-miss-competitive relative to \( Q \).
If \( Q \) incurs \( x \) misses, then \( P \) incurs at most \( 3 \cdot x + 4 \) misses.
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**Best:** \( P \) is \((1, 0)\)-miss-competitive relative to \( Q \).
Example – Relative Miss-Competitiveness

**P** is \((3, 4)\)-miss-competitive relative to **Q**.
If **Q** incurs \(x\) misses, then **P** incurs at most \(3 \cdot x + 4\) misses.

**Best:** **P** is \((1, 0)\)-miss-competitive relative to **Q**.

**Worst:** **P** is not-miss-competitive (or \(\infty\)-miss-competitive) relative to **Q**.
Example – Relative Hit-Competitiveness

\( P \) is \((\frac{2}{3}, 3)\)-hit-competitive relative to \( Q \).
If \( Q \) has \( x \) hits, then \( P \) has at least \( \frac{2}{3} \cdot x - 3 \) hits.
Example – Relative Hit-Competitiveness

\( P \) is \( \left( \frac{2}{3}, 3 \right) \)-hit-competitive relative to \( Q \).
If \( Q \) has \( x \) hits, then \( P \) has at least \( \frac{2}{3} \cdot x - 3 \) hits.

**Best:** \( P \) is \( (1, 0) \)-hit-competitive relative to \( Q \).
Equivalent to \( (1, 0) \)-miss-competitiveness.

**Worst:** \( P \) is \( (0, 0) \)-hit-competitive relative to \( Q \).
Analogue to \( \infty \)-miss-competitiveness.
Example – Relative Hit-Competitiveness

P is \((\frac{2}{3}, 3)\)-hit-competitive relative to Q.
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Analogue to \(\infty\)-miss-competitiveness.
Local Guarantees: (1, 0)-Competitiveness

Let $P$ be $(1, 0)$-competitive relative to $Q$:

$$m_P(p, s) \leq 1 \cdot m_Q(q, s) + 0$$

$$\iff m_P(p, s) \leq m_Q(q, s)$$
Local Guarantees: (1, 0)-Competitiveness

Let $P$ be (1, 0)-competitive relative to $Q$:

$$m_P(p, s) \leq 1 \cdot m_Q(q, s) + 0$$

$$\iff m_P(p, s) \leq m_Q(q, s)$$

1. If $Q$ “hits”, so does $P$, and
2. if $P$ “misses”, so does $Q$. 
Local Guarantees: (1, 0)-Competitiveness

Let $P$ be (1, 0)-competitive relative to $Q$:

$$m_P(p, s) \leq 1 \cdot m_Q(q, s) + 0$$

$$\iff m_P(p, s) \leq m_Q(q, s)$$

1. If $Q$ “hits”, so does $P$, and
2. if $P$ “misses”, so does $Q$.

As a consequence,

1. a *must*-analysis for $Q$ is also a *must*-analysis for $P$, and
2. a *may*-analysis for $P$ is also a *may*-analysis for $Q$. 
Global Guarantees: \((k, c)\)-Competitiveness

**Given:** Global guarantees for policy \(Q\).

**Wanted:** Global guarantees for policy \(P\).
Global Guarantees: \((k, c)\)-Competitiveness

Given: Global guarantees for policy \(Q\).

Wanted: Global guarantees for policy \(P\).

1. Determine competitiveness of policy \(P\) relative to policy \(Q\).

\[
m_P \leq k \cdot m_Q + c
\]
Global Guarantees: \((k, c)\)-Competitiveness

Given: Global guarantees for policy \(Q\).
Wanted: Global guarantees for policy \(P\).

1. Determine competitiveness of policy \(P\) relative to policy \(Q\).
   \[ m_P \leq k \cdot m_Q + c \]

2. Compute global guarantee for task \(T\) under policy \(Q\).
   \[ m_Q(T) \]
Global Guarantees: \((k, c)\)-Competitiveness

**Given:** Global guarantees for policy \(Q\).

**Wanted:** Global guarantees for policy \(P\).

1. Determine competitiveness of policy \(P\) relative to policy \(Q\).
   \[
   m_p \leq k \cdot m_Q + c
   \]

2. Compute global guarantee for task \(T\) under policy \(Q\).
   \[
   m_Q(T)
   \]

3. Calculate global guarantee on the number of misses for \(P\) using the global guarantee for \(Q\) and the competitiveness results of \(P\) relative to \(Q\).
   \[
   m_p \leq k \cdot m_Q + c
   \]
   
   \[
   m_Q(T) = m_P(T)
   \]
Relative Competitiveness: Automatic Computation

$P$ and $Q$ (here: FIFO and LRU) induce transition system:

$$\begin{align*}
\text{Legend} & \\
\text{[abcd]}_{\text{FIFO}} & \quad \text{Cache-set state} \\
\text{Memory access} & \quad (h,m), \ldots \\
\text{Misses in pairs of cache-set states} & \\
\text{Jan Reineke} & \quad \text{Caches in WCET Analysis} \\
\text{ARTIST Summer School 2009} & \quad 39 / 51
\end{align*}$$

Competitive miss ratio $= \text{maximum ratio of misses in policy } P \text{ to misses in policy } Q \text{ in transition system}$
Transition System is $\infty$ Large

Problem: The induced transition system is $\infty$ large.

Observation: Only the relative positions of elements matter:

\[
\begin{align*}
[abc]_{LRU}, [bde]_{FIFO} & \approx \quad [fgl]_{LRU}, [ghm]_{FIFO} \\
c (h, m) & \quad \quad \quad l (h, m)
\end{align*}
\]

Solution: Construct finite quotient transition system.
-Equivalent States in Running Example

\[ eabc \] \_FIFO, \[ eabc \] \_LRU

\[ abcd \] \_FIFO, \[ abcd \] \_LRU

\[ eabc \] \_FIFO, \[ ceab \] \_LRU

\[ abcd \] \_FIFO, \[ dabc \] \_LRU

\[ eabc \] \_FIFO, \[ ceda \] \_LRU

\[ abcd \] \_FIFO, \[ edab \] \_LRU

\[ deab \] \_FIFO, \[ deab \] \_LRU

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Caches in WCET Analysis
ARTIST Summer School 2009
Merging $\approx$-equivalent states yields a finite quotient transition system:

\[
\begin{align*}
[abcd]_{\text{FIFO}}, [abcd]_{\text{LRU}} \\
(h, h) \\
\text{FIFO, LRU} \\
\end{align*}
\]

\[
\begin{align*}
[abcd]_{\text{FIFO}}, [dabc]_{\text{LRU}} \\
(h, h) \\
\text{LRU} \\
\end{align*}
\]

\[
\begin{align*}
[eabc]_{\text{FIFO}}, [ceda]_{\text{LRU}} \\
(h, h) \\
\text{LRU} \\
\end{align*}
\]

\[
\begin{align*}
[eabc]_{\text{FIFO}}, [edab]_{\text{LRU}} \\
(h, m) \\
\text{LRU} \\
\end{align*}
\]

\[
\begin{align*}
[eabc]_{\text{FIFO}}, [edab]_{\text{LRU}} \\
(m, h) \\
\text{LRU} \\
\end{align*}
\]
Competitive Ratio = Maximum Cycle Ratio

Competitive miss ratio =
maximum ratio of misses in policy P to misses in policy Q

\[(0, 0) \rightarrow (1, 1) \rightarrow (0, 0) \rightarrow (1, 1) \rightarrow (0, 1) \rightarrow (1, 0) \rightarrow (0, 0) \rightarrow (1, 1) \rightarrow (0, 1) \rightarrow (1, 0) \rightarrow (0, 0) \rightarrow \ldots\]
Competitive Ratio = Maximum Cycle Ratio

Competitive miss ratio =

maximum ratio of misses in policy P to misses in policy Q

Maximum cycle ratio = \( \frac{0+1+1}{0+1+0} = 2 \)
Tool Implementation

- Implemented in Java, called Relacs
- Interface for replacement policies
- Fully automatic
- Provides example sequences for competitive ratio and constant
- Analysis usually practically feasible up to associativity 8
  - limited by memory consumption
  - depends on similarity of replacement policies

Online version:

http://rw4.cs.uni-sb.de/~reineke/relacs
Generalizations

Identified patterns and proved generalizations by hand. Aided by example sequences generated by tool.
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Identified patterns and proved generalizations by hand. Aided by example sequences generated by tool.

Previously unknown facts:

\[ \text{PLRU}(k) \text{ is } (1, 0) \text{ comp. rel. to } \text{LRU}(1 + \log_2 k), \]
\[ \rightarrow \text{LRU}-must\text{-analysis can be used for PLRU} \]
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\[
\text{FIFO}(k) \text{ is } \left(\frac{1}{2}, \frac{k-1}{2}\right) \text{ hit-comp. rel. to } \text{LRU}(k), \text{ whereas}
\]

\[
\text{LRU}(k) \text{ is } (0, 0) \text{ hit-comp. rel. to } \text{FIFO}(k), \text{ but}
\]
Generalizations

Identified patterns and proved generalizations by hand.
Aided by example sequences generated by tool.

Previously unknown facts:

PLRU\((k)\) is \((1, 0)\) comp. rel. to LRU\((1 + \log_2 k)\),
\rightarrow LRU-\emph{must}-analysis can be used for PLRU

FIFO\((k)\) is \(\left(\frac{1}{2}, \frac{k-1}{2}\right)\) hit-comp. rel. to LRU\((k)\), whereas
LRU\((k)\) is \((0, 0)\) hit-comp. rel. to FIFO\((k)\), but

LRU\((2k - 1)\) is \((1, 0)\) comp. rel. to FIFO\((k)\), and
LRU\((2k - 2)\) is \((1, 0)\) comp. rel. to MRU\((k)\).
\rightarrow LRU-\emph{may}-analysis can be used for FIFO and MRU
\rightarrow optimal with respect to predictability metric Evict
Generalizations

Identified patterns and proved generalizations by hand. Aided by example sequences generated by tool.

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\[
\text{PLRU}(k) \text{ is } (1, 0) \text{ comp. rel. to } \text{LRU}(1 + \log_2 k), \\
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\[
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\]

\[
\text{LRU}(k) \text{ is } (0, 0) \text{ hit-comp. rel. to FIFO}(k), \text{ but}
\]

\[
\text{LRU}(2k - 1) \text{ is } (1, 0) \text{ comp. rel. to FIFO}(k), \text{ and}
\]

\[
\text{LRU}(2k - 2) \text{ is } (1, 0) \text{ comp. rel. to MRU}(k),
\rightarrow \text{LRU-}may\text{-analysis can be used for FIFO and MRU}
\rightarrow \text{optimal with respect to predictability metric Evict}
\]

FIFO-\textit{may}-analysis used in the analysis of the branch target buffer of the \textsc{Motorola PowerPC 56X}. 
Outline

1 Caches

2 Cache Analysis for Least-Recently-Used

3 Beyond Least-Recently-Used
   - Predictability Metrics
   - Relative Competitiveness
   - Sensitivity – Caches and Measurement-Based Timing Analysis

4 Summary
Measurement-Based Timing Analysis

- Run program on a number of inputs and initial states.
- Combine measurements for basic blocks to obtain WCET estimation.
- Sensitivity Analysis demonstrates this approach may be dramatically wrong.
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- Combine measurements for basic blocks to obtain WCET estimation.
- Sensitivity Analysis demonstrates this approach may be dramatically wrong.
Influence of Initial Cache State

variation due to initial cache state

BCET \hspace{2cm} WCET upper bound execution time

Definition (Miss sensitivity)

Policy $P$ is $(k, c)$-miss-sensitive if

$$m_P(p, s) \leq k \cdot m_P(p', s) + c$$

for all access sequences $s \in M^*$ and cache-set states $p, p' \in C^P$. 
### Sensitivity Results

<table>
<thead>
<tr>
<th>Policy</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
</tr>
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<tbody>
<tr>
<td>LRU</td>
<td>1,2</td>
<td>1,3</td>
<td>1,4</td>
<td>1,5</td>
<td>1,6</td>
<td>1,7</td>
<td>1,8</td>
</tr>
<tr>
<td>FIFO</td>
<td>2,2</td>
<td>3,3</td>
<td>4,4</td>
<td>5,5</td>
<td>6,6</td>
<td>7,7</td>
<td>8,8</td>
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<td>PLRU</td>
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<td></td>
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<td></td>
<td></td>
<td></td>
<td>∞</td>
</tr>
<tr>
<td>MRU</td>
<td>1,2</td>
<td>3,4</td>
<td>5,6</td>
<td>7,8</td>
<td>MEM</td>
<td>MEM</td>
<td>MEM</td>
</tr>
</tbody>
</table>

- LRU is optimal. Performance varies in the least possible way.
- For FIFO, PLRU, and MRU the number of misses may vary strongly.
- Case study based on simple model of execution time by Hennessy and Patterson (2003): WCET may be 3 times higher than a measured execution time for 4-way FIFO.
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...efficiently represents sets of cache states by bounding the age of memory blocks from above and below.

...requires context-sensitivity for precision.
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→ LRU is the most predictable policy.
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Thank you for your attention!
**Most-Recently-Used – MRU**

MRU-bits record whether line was recently used.

```
[abcd]_0101  ⇒ b,d

[ebcd]_1101  ⇒ e,b,d

[ebcd]_0010  ⇒ c
```

→ Never converges
Hit on a “rejuvenates” neighborhood; “saves” b from eviction.
May- and Must-Information

\[ May^P(s) := \bigcup_{p \in C^P} CC^P(update^P(p, s)) \]

\[ Must^P(s) := \bigcap_{p \in C^P} CC^P(update^P(p, s)) \]

\[ may^P(n) := \left| May^P(s) \right| \text{, where } s \in S^\neq \subsetneq M^*, |s| = n \]

\[ must^P(n) := \left| Must^P(s) \right| \text{, where } s \in S^\neq \subsetneq M^*, |s| = n \]

\( S^\neq \): set of finite access sequences with pairwise different accesses
Definitions of Metrics

\[ Evict^P := \min \left\{ n \mid may^P(n) \leq n \right\}, \]
\[ Fill^P := \min \left\{ n \mid must^P(n) = k \right\}, \]

where \( k \) is \( P \)'s associativity.
Let $P(k)$ be $(1, 0)$-miss-competitive relative to policy $Q(l)$, then

(i) $\text{Evict}^P(k) \geq \text{Evict}^Q(l)$,
(ii) $\text{mls}^P(k) \geq \text{mls}^Q(l)$. 
Let $l$ be the smallest associativity, such that $LRU(l)$ is $(1, 0)$-miss-competitive relative to $P(k)$. Then

$$\text{Alt-Evict}^P(k) = l.$$ 

Let $l$ be the greatest associativity, such that $P(k)$ is $(1, 0)$-miss-competitive relative to $LRU(l)$. Then

$$\text{Alt-mls}^P(k) = l.$$
Size of Transition System

\[ 2^{l+l'} \cdot \sum_{i=0}^{k} \binom{k}{i} \cdot \sum_{i'=0}^{k'} \binom{k'}{i'} \cdot \sum_{j=0}^{\min\{i,i'\}} \binom{i}{j} \binom{i'}{j} j! \]

- Status bits of \( P \) and \( Q \)
- Non-empty lines in \( P \)
- Non-empty lines in \( Q \)
- Number of overlappings in non-empty lines

\[ \min\{k,k'\} \sum_{j=0}^{\infty} \frac{1}{j!} = e \cdot k! \cdot k'! \]

This can be bounded by

\[ 2^{l+l'+k+k'} \leq |(C_k^l \times C_{k'}^{l'})| \approx \leq 2^{l+l'+k+k'} \cdot e \cdot k! \cdot k'! \]

bound on number of overlappings
Compatible States

\[ i^P = [\bot \bot \bot \bot]_P \approx i^Q = [\bot \bot \bot \bot]_Q \]

\[ update_P(i^P, s) \approx update_Q(i^Q, s) \]
Let $P$ be $(1, 0)$-competitive relative to $Q$, then

$$m_P(p, \langle x \rangle) = 1 \quad \Rightarrow \quad m_Q(q, \langle x \rangle) = 1$$
(1, 0)-Competitiveness and May/Must-Analyses

\[ \forall p \in P : m_P(p, \langle x \rangle) = 1 \quad \Rightarrow \quad \forall q \in Q : m_Q(q, \langle x \rangle) = 1 \]
Case Study: Impact of Sensitivity

- Simple model of execution time from Hennessy & Patterson (2003)
- $CPI_{hit} = \text{Cycles per instruction assuming cache hits only}$
- $\frac{\text{Memory accesses}}{\text{Instruction}}$ including instruction and data fetches

\[
\frac{T_{\text{wc}}}{T_{\text{meas}}} = \frac{\text{CPI}_{hit} + \frac{\text{Memory accesses}}{\text{Instruction}} \times \text{Miss rate}_{wc} \times \text{Miss penalty}}{\text{CPI}_{hit} + \frac{\text{Memory accesses}}{\text{Instruction}} \times \text{Miss rate}_{meas} \times \text{Miss penalty}}
\]

\[
= \frac{1.5 + 1.2 \times 0.20 \times 50}{1.5 + 1.2 \times 0.05 \times 50} = \frac{13.5}{4.5} = 3
\]