

# Evaluation and Validation

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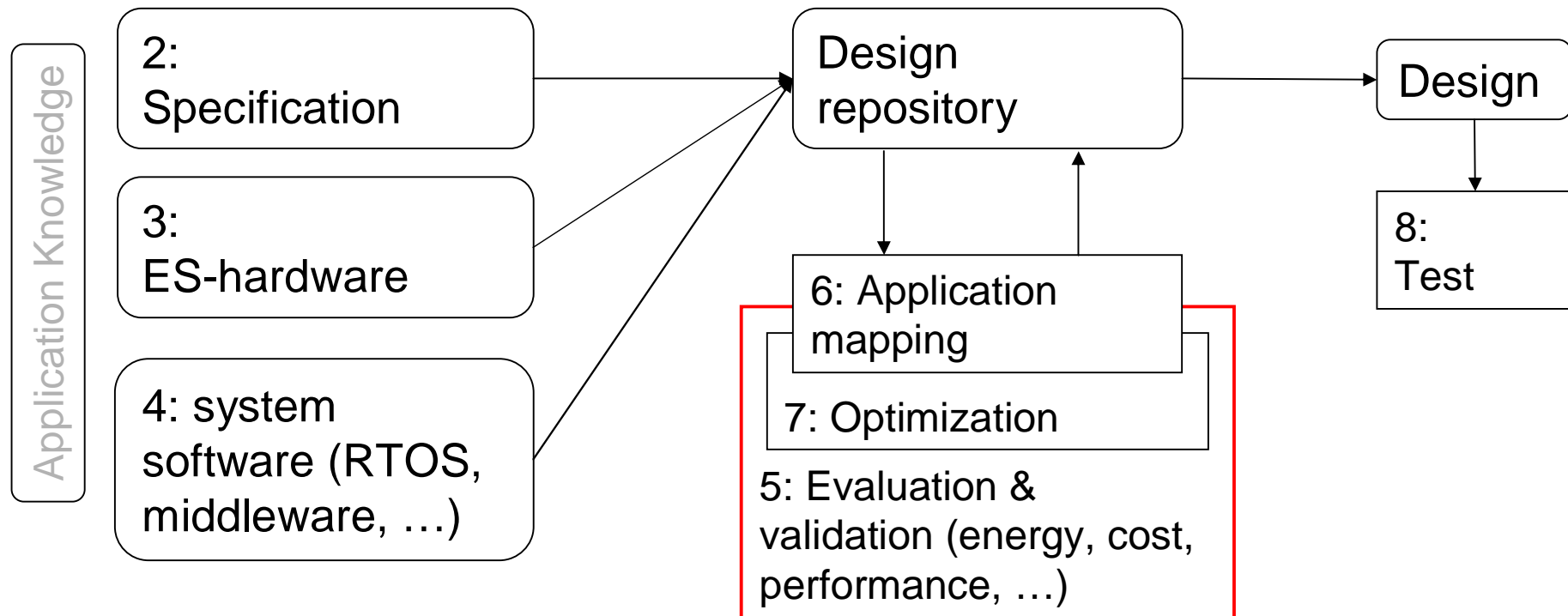


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# Structure of this course



Numbers denote sequence of chapters

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# Validation and Evaluation

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**Definition:** Validation is the process of checking whether or not a certain (possibly partial) design is appropriate for its purpose, meets all constraints and will perform as expected (yes/no decision).

**Definition:** Validation with mathematical rigor is called (formal) verification.

**Definition:** Evaluation is the process of computing quantitative information of some key characteristics of a certain (possibly partial) design.

# How to evaluate designs according to multiple criteria?

Many different criteria are relevant for evaluating designs:

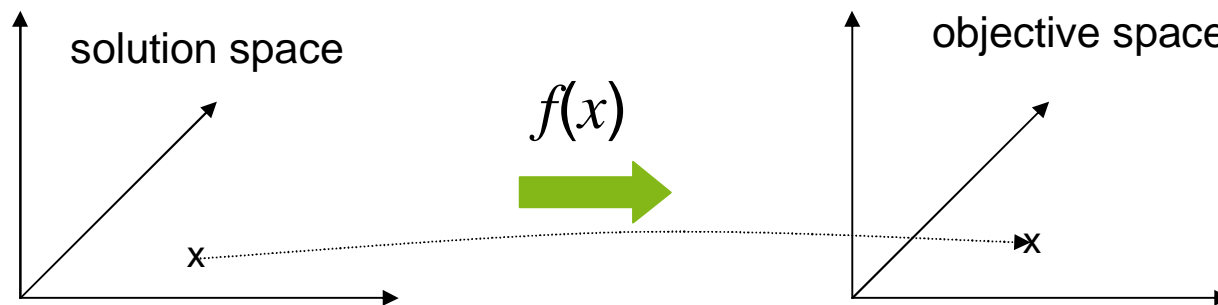
- Average & worst case delay
- power/energy consumption
- thermal behavior
- reliability, safety, security
- cost, size
- weight
- EMC characteristics
- radiation hardness, environmental friendliness, ..



How to compare different designs?  
(Some designs are “better” than others)

# Definitions

- Let  $X$ :  $m$ -dimensional **solution space** for the design problem.  
Example: dimensions correspond to # of processors, size of memories, type and width of busses etc.
- Let  $F$ :  $n$ -dimensional **objective space** for the design problem.  
Example: dimensions correspond to average and worst case delay, power/energy consumption, size, weight, reliability, ...
- Let  $f(x) = (f_1(x), \dots, f_n(x))$  where  $x \in X$  be an **objective function**.  
We assume that we are using  $f(x)$  for evaluating designs.



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# Pareto points

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- We assume that, for each objective, an order  $<$  and the corresponding order  $\leq$  are defined.

- **Definition:**

Vector  $u=(u_1,\dots,u_n)\in F$  **dominates** vector  $v=(v_1,\dots,v_n)\in F$

$\Leftrightarrow$

$u$  is “better” than  $v$  with respect to one objective and not worse than  $v$  with respect to all other objectives:

$$\forall i \in \{1,\dots,n\} : u_i \leq v_i \wedge$$

$$\exists i \in \{1,\dots,n\} : u_i < v_i$$

- **Definition:**

Vector  $u\in F$  is **indifferent** with respect to vector  $v\in F$

$\Leftrightarrow$  neither  $u$  dominates  $v$  nor  $v$  dominates  $u$

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# Pareto points

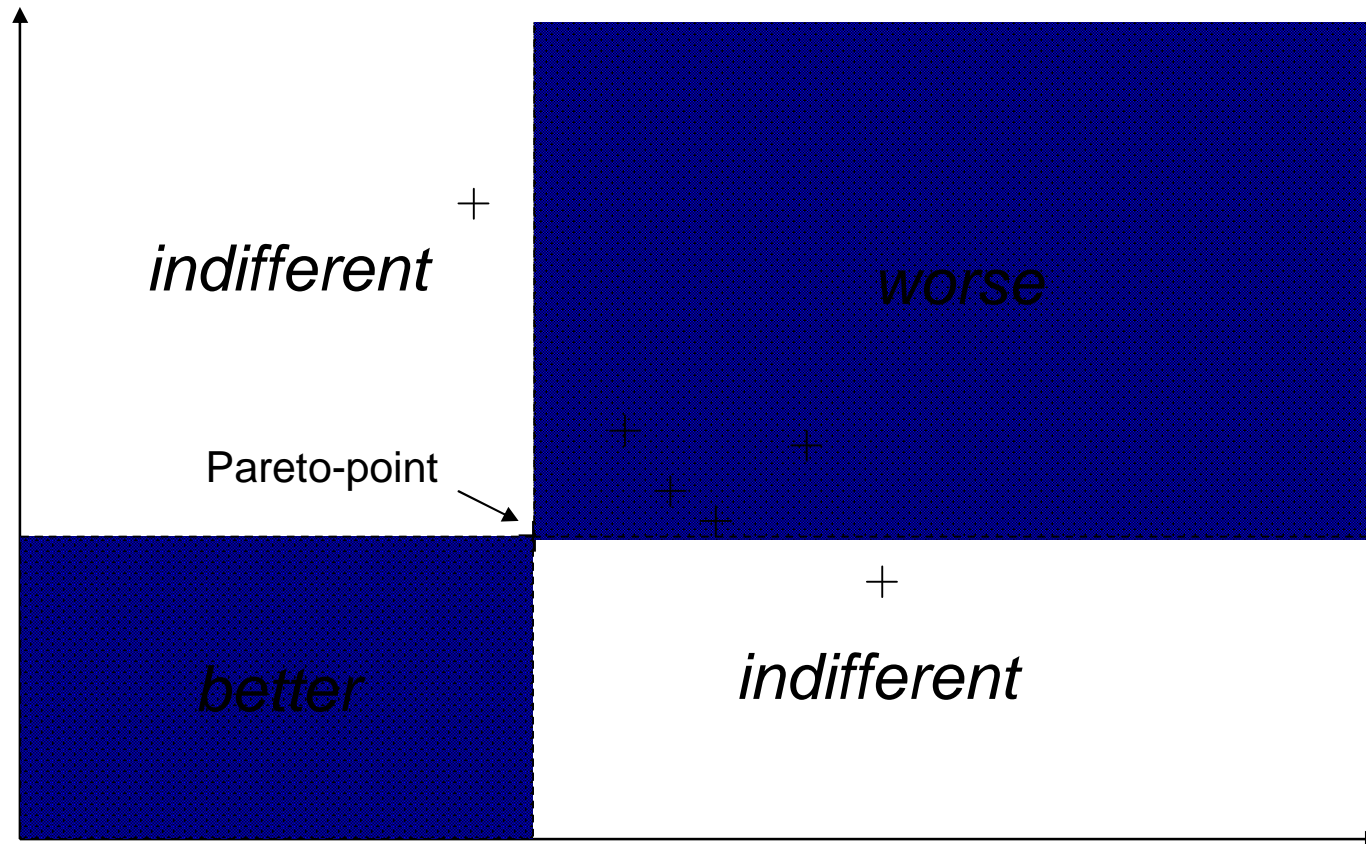
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- A solution  $x \in X$  is called **Pareto-optimal** with respect to  $X$   
 $\Leftrightarrow$  there is no solution  $y \in X$  such that  $u=f(x)$  is dominated by  $v=f(y)$ .  $x$  is a **Pareto point**.
- **Definition:** Let  $S \subseteq F$  be a subset of solutions.  
 $v \in F$  is called a **non-dominated solution** with respect to  $S$   
 $\Leftrightarrow v$  is not dominated by any element  $\in S$ .
- $v$  is called **Pareto-optimal**  
 $\Leftrightarrow v$  is non-dominated with respect to all solutions  $F$ .
- A **Pareto-set** is the set of all Pareto-optimal solutions

Pareto-sets define a **Pareto-front**  
(boundary of dominated subspace)

# Pareto Point

Objective 1  
(e.g. energy  
consumption)



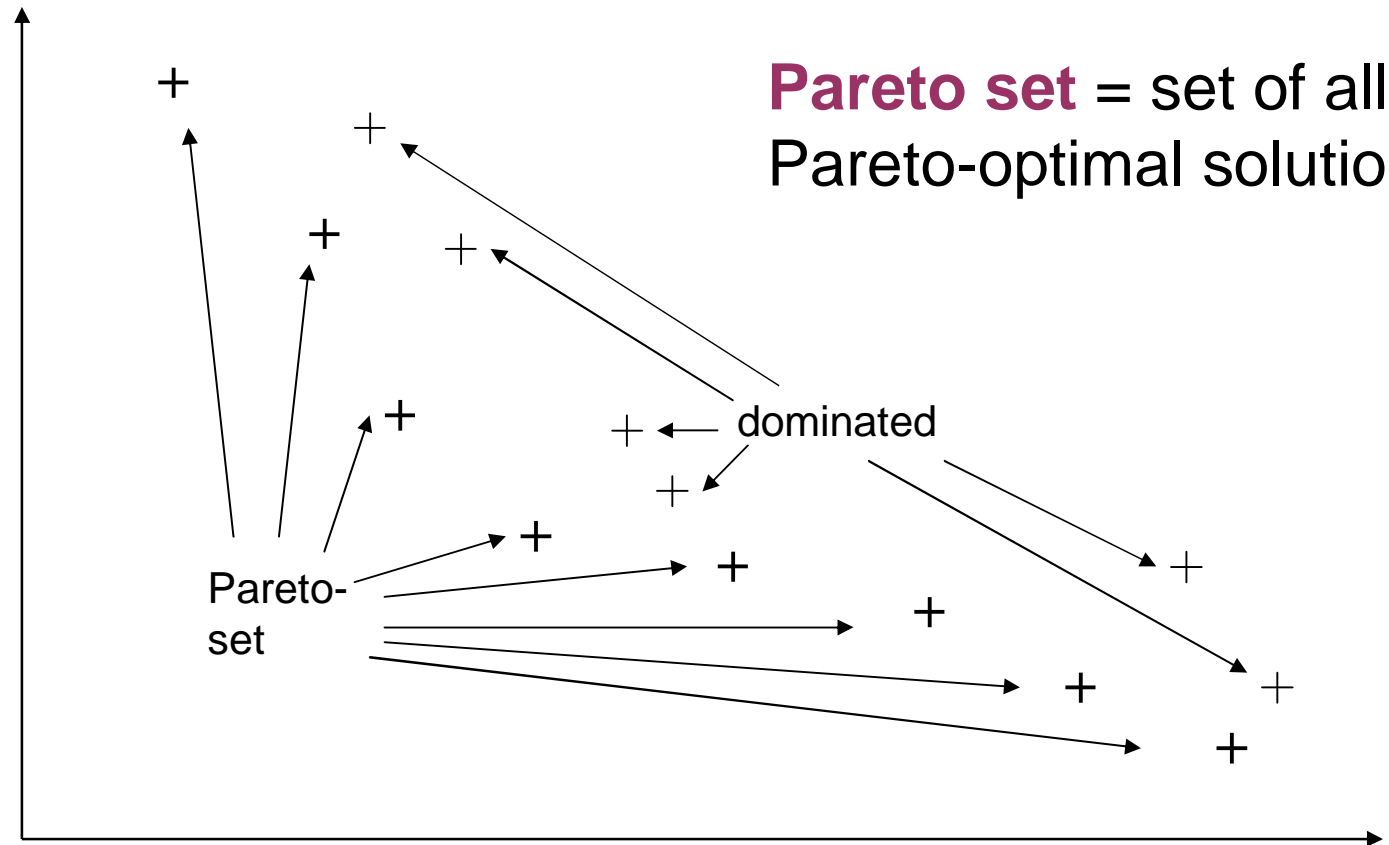
Objective 2  
(e.g. run time)

(Assuming *minimization* of objectives)



# Pareto Set

Objective 1  
(e.g. energy  
consumption)

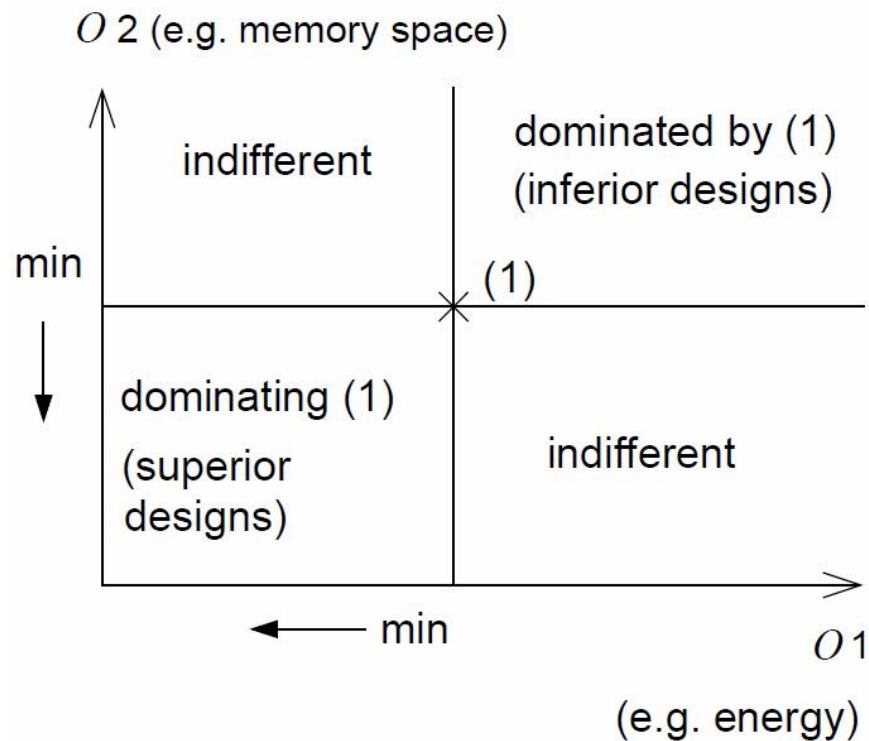


Objective 2  
(e.g. run time)

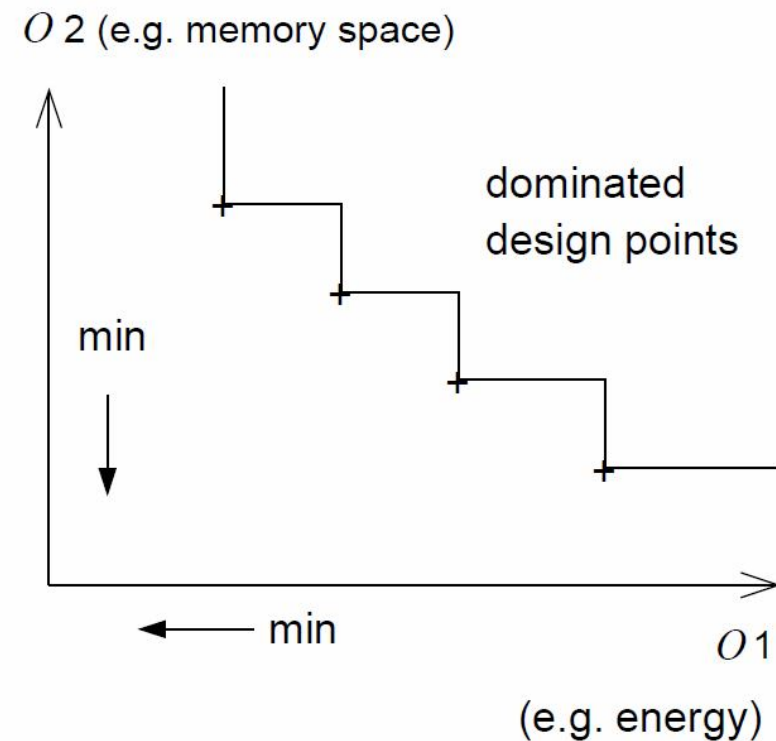
(Assuming *minimization* of objectives)

# One more time ...

## Pareto point



## Pareto front



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# Design space evaluation

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**Design space evaluation** (DSE) based on Pareto-points is the process of finding and returning a set of Pareto-optimal designs to the user, enabling the user to select the most appropriate design.

# How to evaluate designs according to multiple criteria?

Many different criteria are relevant for evaluating designs:

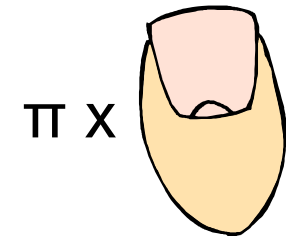
- Average & worst case delay
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How to compare different designs?  
(Some designs are “better” than others)

# Average delays (execution times)

- **Estimated** average execution times :  
Difficult to generate sufficiently precise estimates;  
Balance between run-time and precision



- **Accurate** average execution times:  
As precise as the input data is.



We need to compute **average** and **worst case** execution times

# Worst/best case execution times



## Requirements on WCET estimates:

- *Safeness*:  $WCET \leq WCET_{EST}$ !
- *Tightness*:  $WCET_{EST} - WCET \rightarrow \text{minimal}$

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# Worst case execution times (2)

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## Complexity:

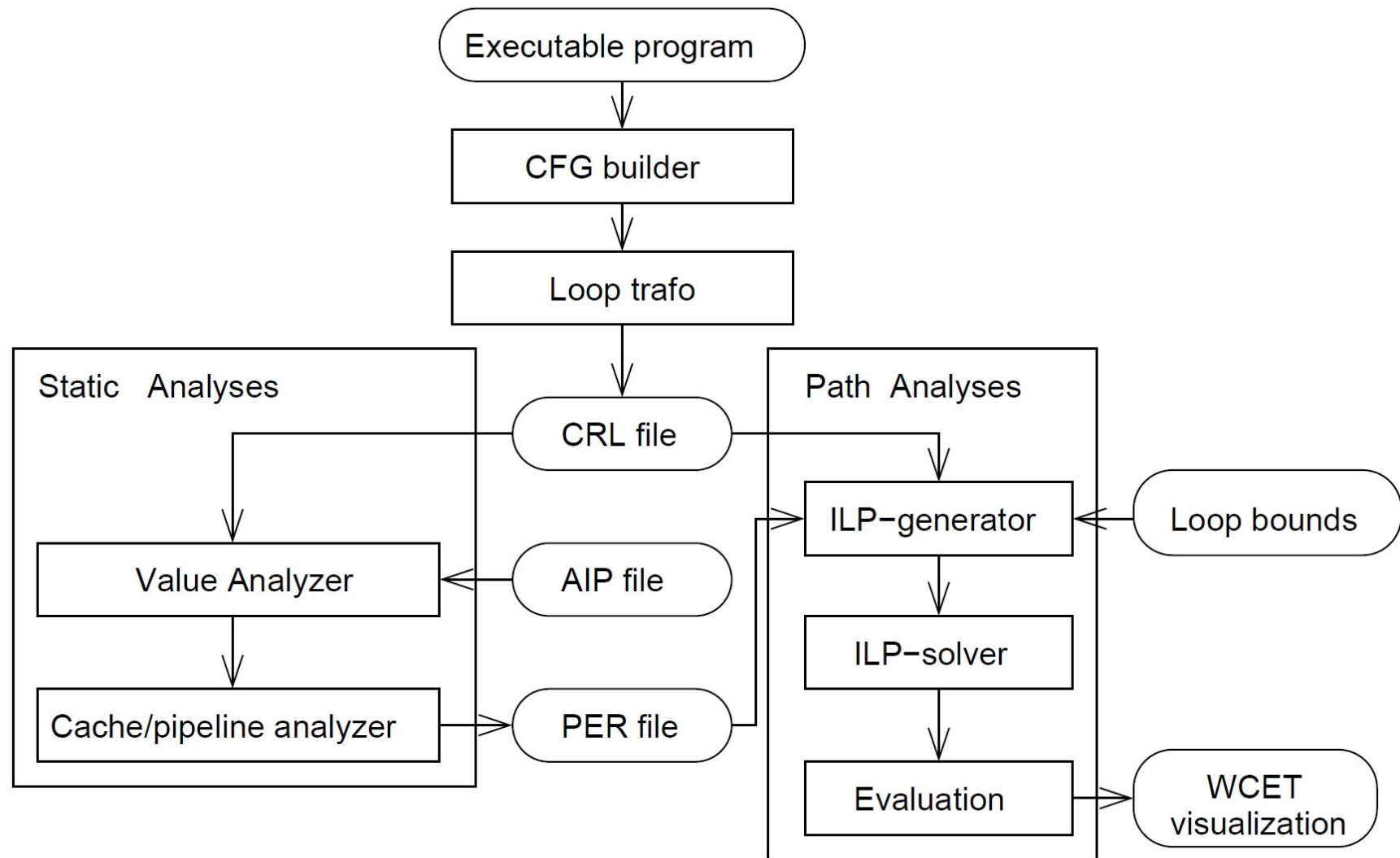
- in the general case: undecidable if a bound exists.
- for restricted programs: simple for “old” architectures, very complex for new architectures with pipelines, caches, interrupts, virtual memory, etc.



## Approaches:

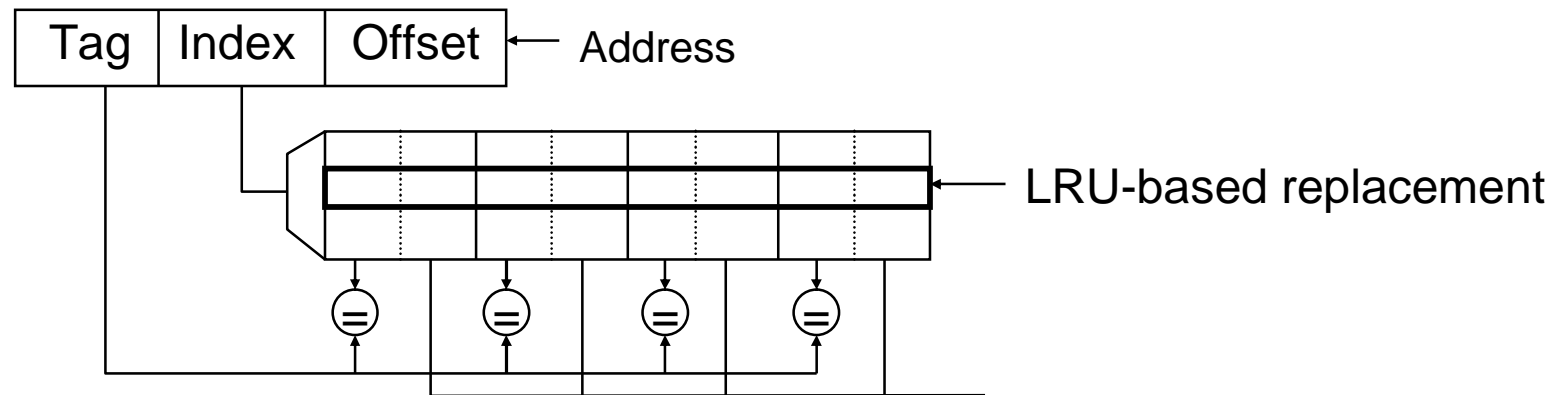
- for hardware: requires detailed timing behavior
- for software: requires availability of machine programs; complex analysis (see, e.g., [www.absint.de](http://www.absint.de))

# WCET estimation: AiT (AbsInt)



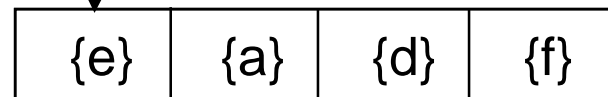


# WCET estimation for caches



Old state

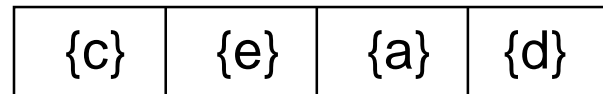
Youngest entry



Reference to c

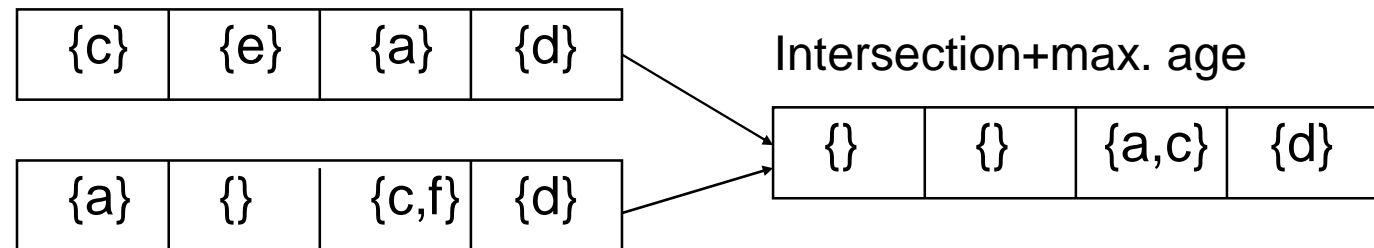
Variables getting older

New state

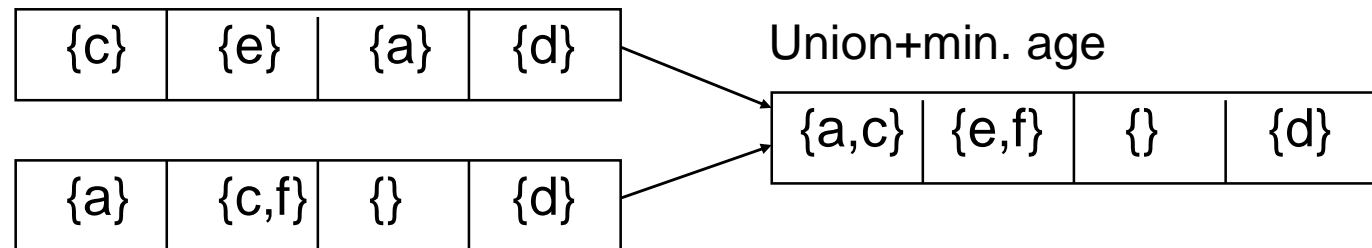


# Behavior at program joins

## Worst case

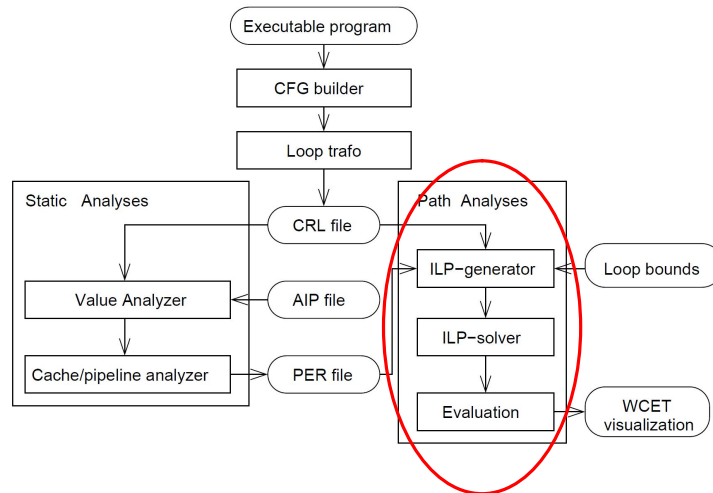


## Best case



➡ Possibly several variables per entry

# ILP model



- Objective function reflects execution time as a function of the execution time of blocks. To be **maximized**.
- Constraints reflect dependencies between blocks.
- Avoids explicit consideration of all paths
- ☞ Called **implicit path enumeration** technique.

# Example (1)

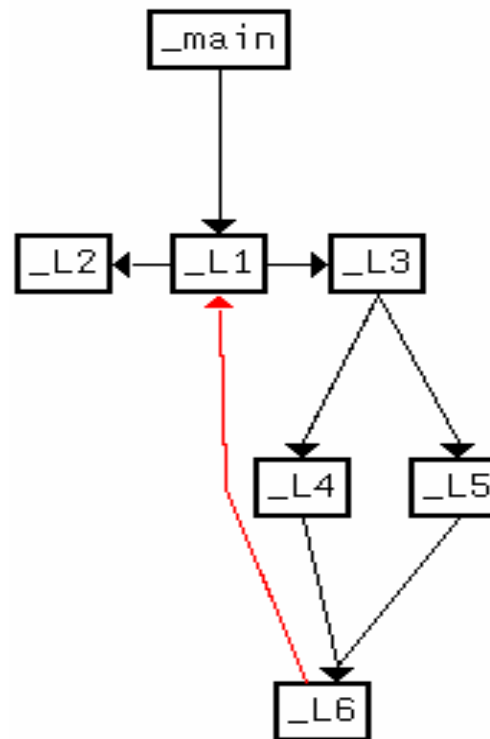
## Program

```
int main()
{
    int i, j = 0;

    _Pragma( "loopbound min
              100 max 100" );
    for ( i = 0; i < 100; i++ ) {
        if ( i < 50 )
            j += i;
        else
            j += ( i * 13 ) % 42;
    }

    return j;
}
```

## CFG



## WCETs of BB (aiT 4 TriCore)

```
_main: 21 cycles
_L1:   27
_L3:    2
_L4:    2
_L5:   20
_L6:   13
_L2:   20
```

# Example (2)

## ILP

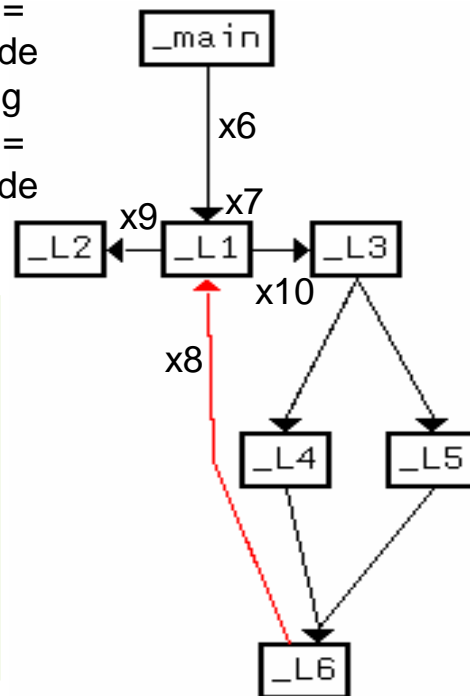
- Virtual start node
- Virtual end node
- Virtual end node per function

Variables:

- 1 variable per node
- 1 variable per edge

Constraints: „Kirchhoff“ equations per node

- Sum of incoming edge variables = flux through node
- Sum of outgoing edge variables = flux through node



`_main`: 21 cycles  
`_L1`: 27  
`_L3`: 2  
`_L4`: 2  
`_L5`: 20  
`_L6`: 13  
`_L2`: 20

```

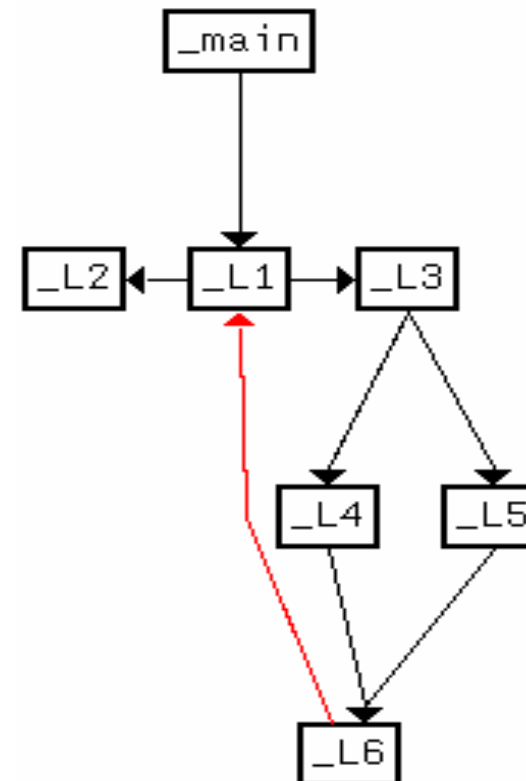
/* Objective function = WCET to be maximized*/
21 x2 + 27 x7 + 2 x11 + 2 x14 + 20 x16 + 13 x18 + 20 x19;
/* CFG Start Constraint */ x0 - x4 = 0;
/* CFG Exit Constraint */ x1 - x5 = 0;
/* Constraint for flow entering function main */
x2 - x4 = 0;
/* Constraint for flow leaving exit node of main */
x3 - x5 = 0;
/* Constraint for flow entering exit node of main */
x3 - x20 = 0;
/* Constraint for flow entering main = flow leaving main */
x2 - x3 = 0;
/* Constraint for flow leaving CFG node _main */
x2 - x6 = 0;
/* Constraint for flow entering CFG node _L1 */
x7 - x8 - x6 = 0;
/* Constraint for flow leaving CFG node _L1 */
x7 - x9 - x10 = 0;
/* Constraint for lower loop bound of _L1 */
x7 - 101 x9 >= 0;
/* Constraint for upper loop bound of _L1 */
x7 - 101 x9 <= 0; ....
  
```

## Example (3)

Value of objective function: 6268

Actual values of the variables:

x2	1
x7	101
x11	100
x14	0
x16	100
x18	100
x19	1
x0	1
x4	1
x1	1
x5	1
x3	1
x20	1
x6	1
x8	100
x9	1
x10	100
x12	100
x13	0
x15	0
x17	100



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# Summary

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## Evaluation and Validation

- In general, multiple objectives
- Pareto optimality
- Design space evaluation (DSE)
- Execution time analysis
  - Trade-off between speed and accuracy
  - Computation of worst case execution times
    - Cache/pipeline analysis
    - ILP model for computing WCET of application from WCET of blocks