

# Adaptive Information Processing: An Effective Way to Improve Perceptron Predictors

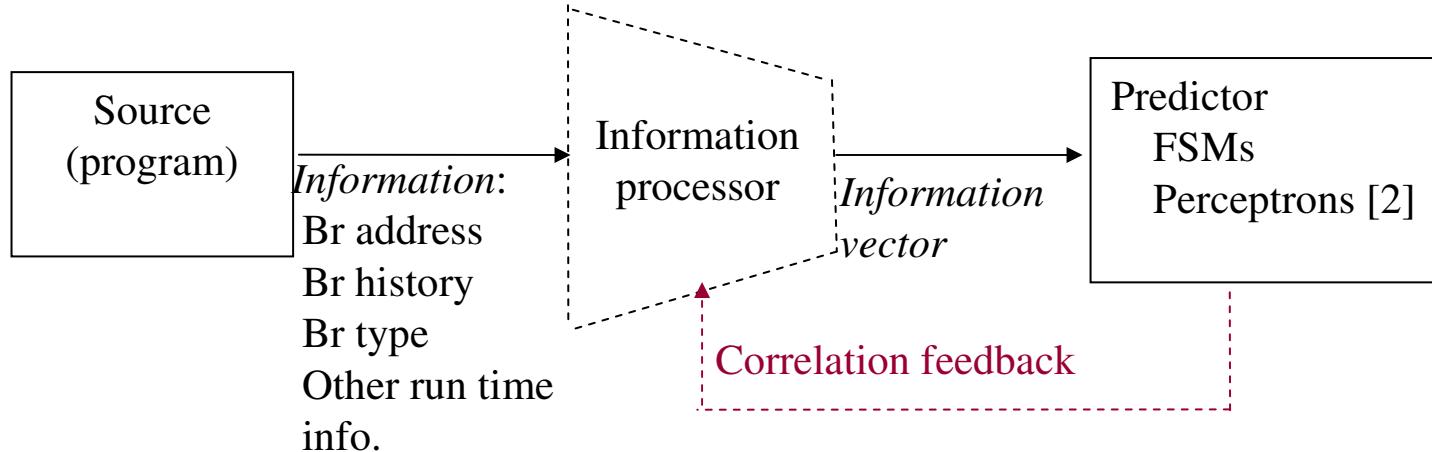
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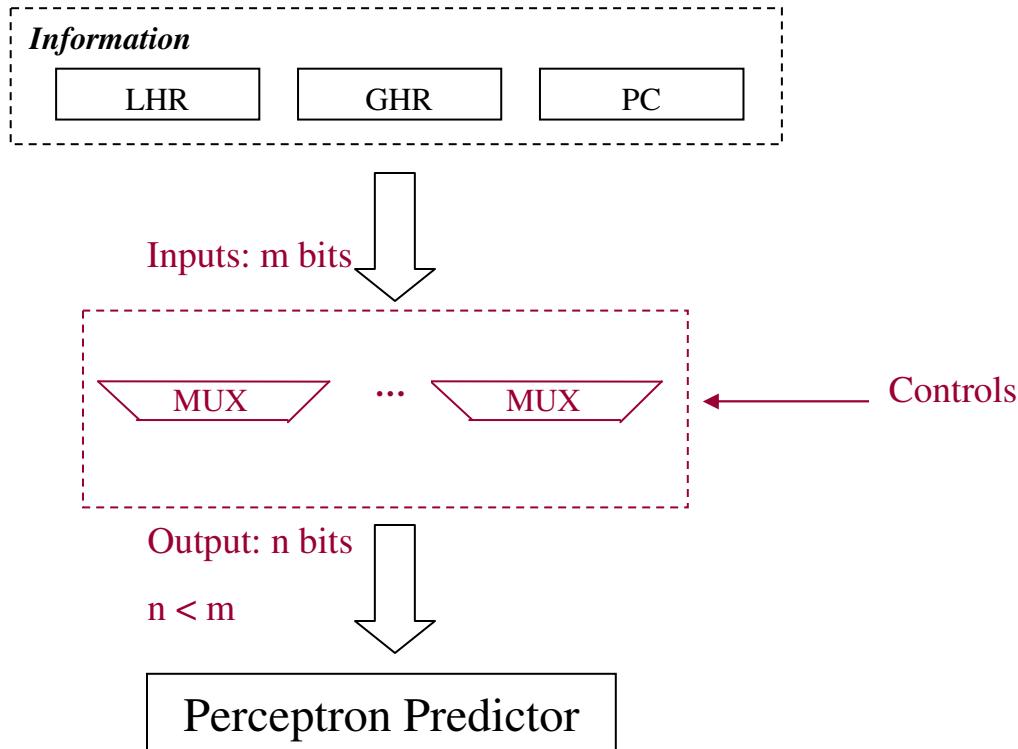
# Information system model



- Information system model by Chen et. al. [ASPLOS-VII].
- Key observations
  - Shortcomings:
    - Fixed information vector while different workloads/branches need different information data.
    - Perceptron weights  $\leftrightarrow$  Correlation
      - Assemble information vector to maximize correlation
- ***Our contribution***
  - Re-assemble the information vector based on correlation (weights)
  - Performed at a coarse grain, so it is not latency critical



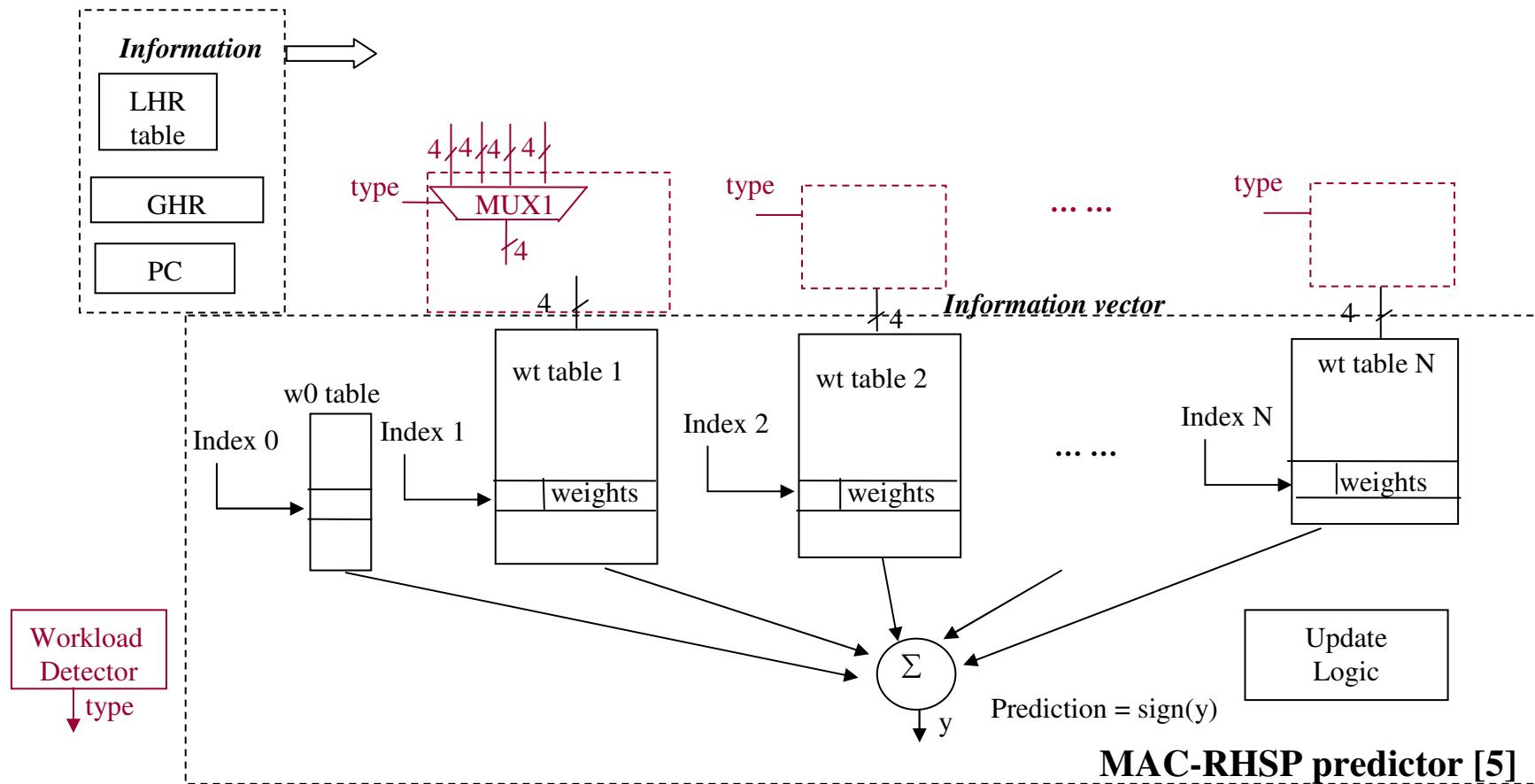
# Adaptive Information Processing



- Profile-directed adaptation
- Correlation-directed adaptation

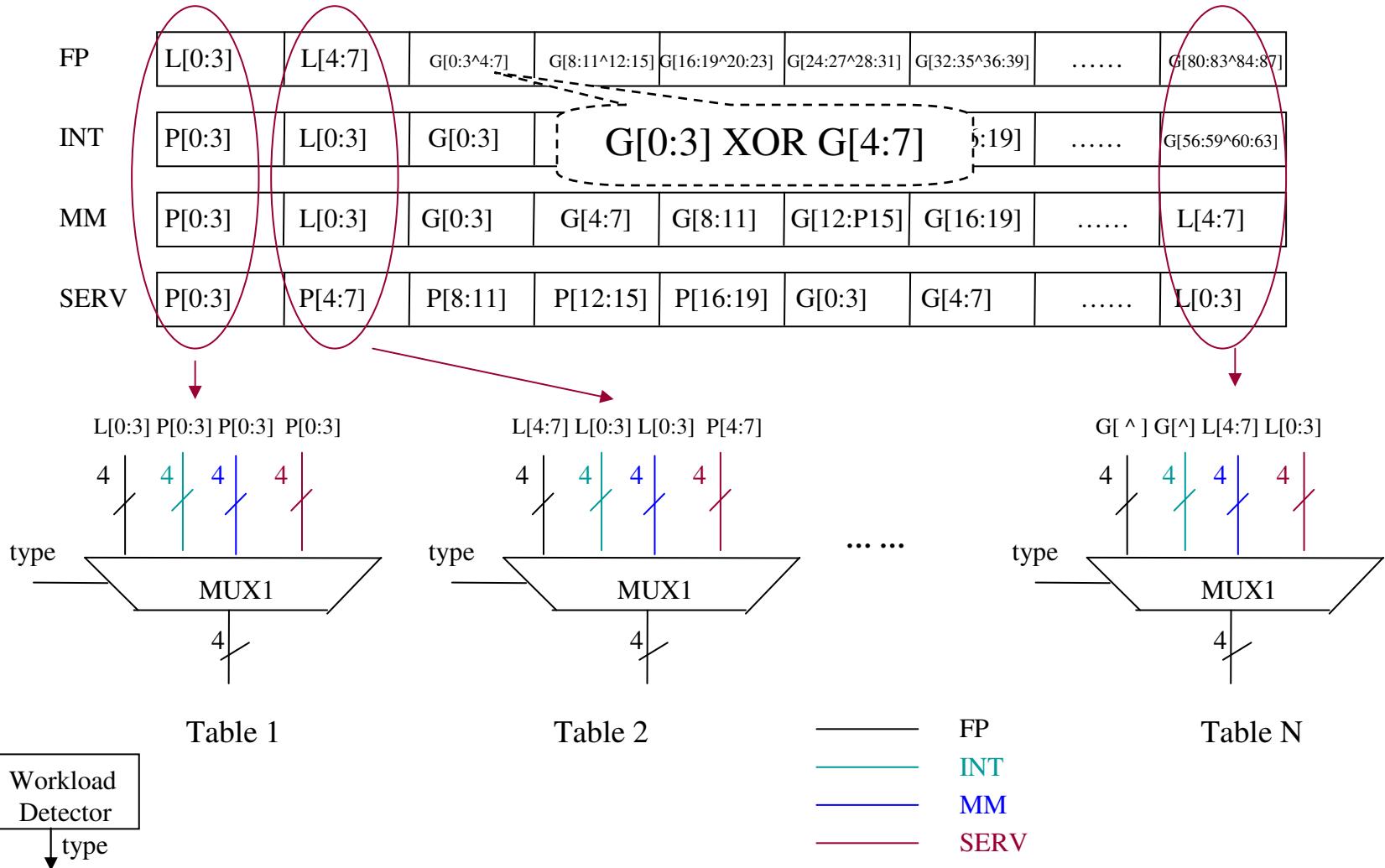


# Profile-directed adaptation



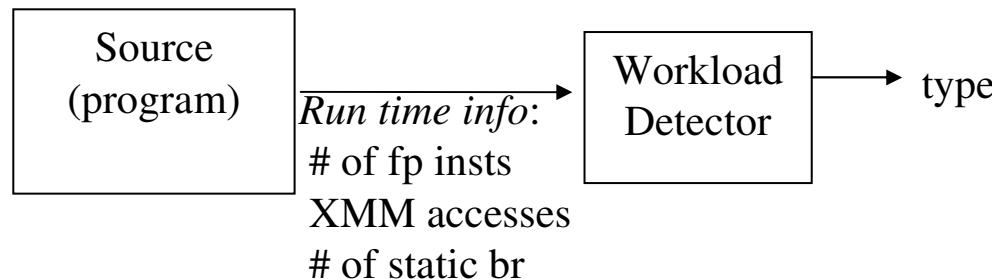


# Profile-directed adaptation





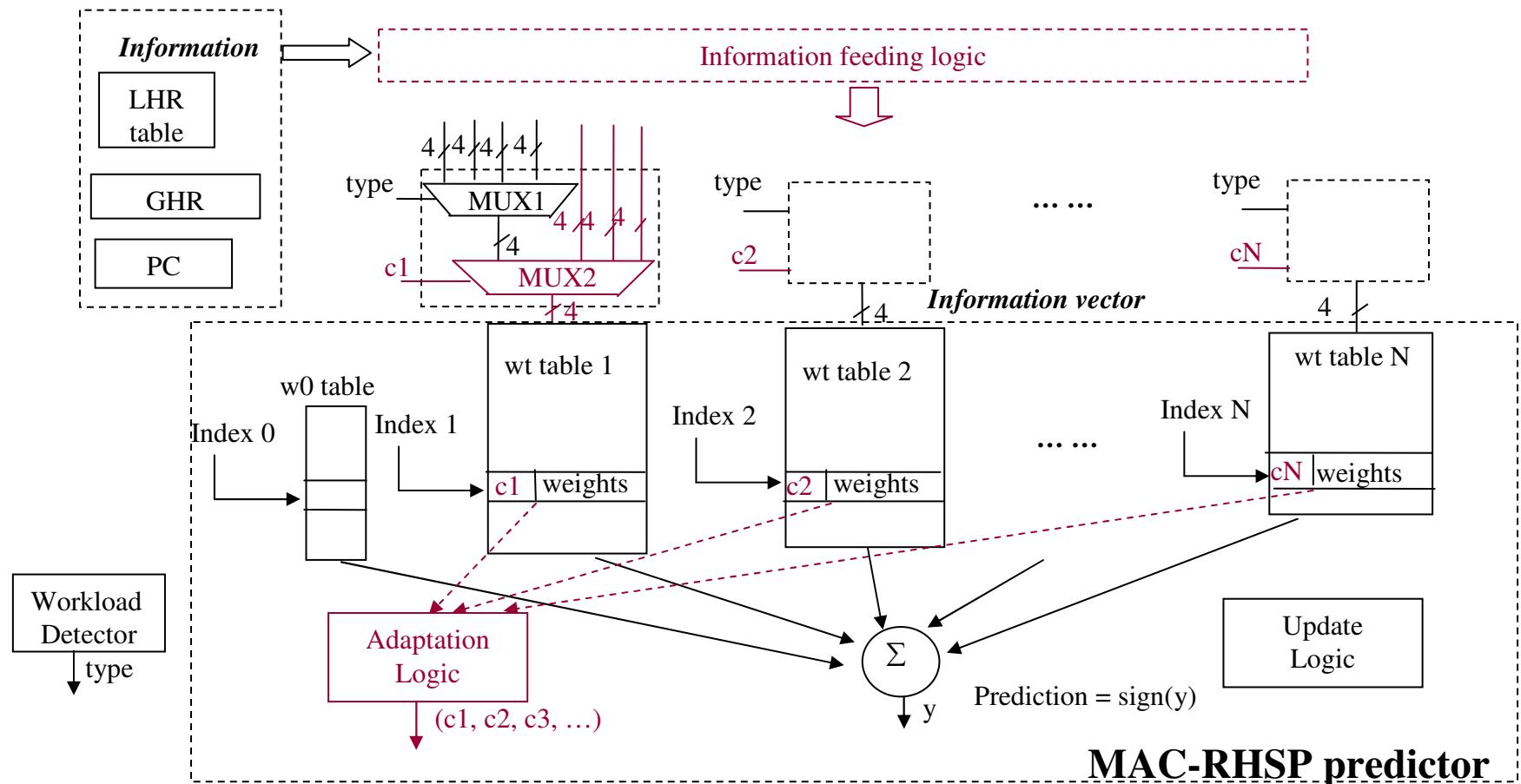
# Workload Detector



- Detection criteria
  - **SERV**: a **large** number of **static branches**
  - **FP**: a **small** number of **static branches**,  
a **high** number of **floating point operation**,  
and a **high** number of instructions using **XMM registers**
  - **MM**: a **medium** number of **static branches**,  
a **medium** number of **floating point operation**,  
and a **medium** number of instructions using **XMM registers**
  - **INT**: default

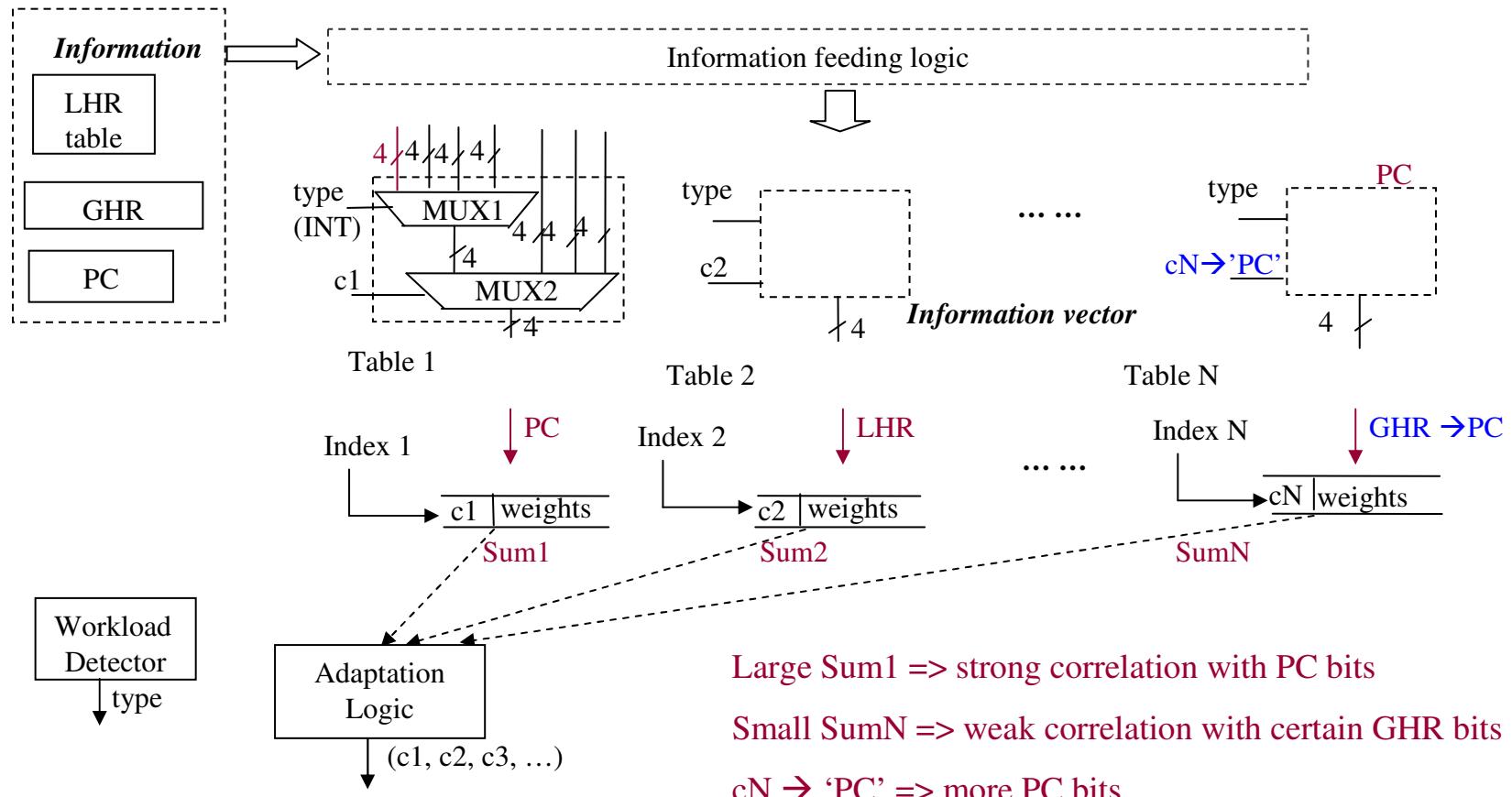


# Correlation-directed adaptation



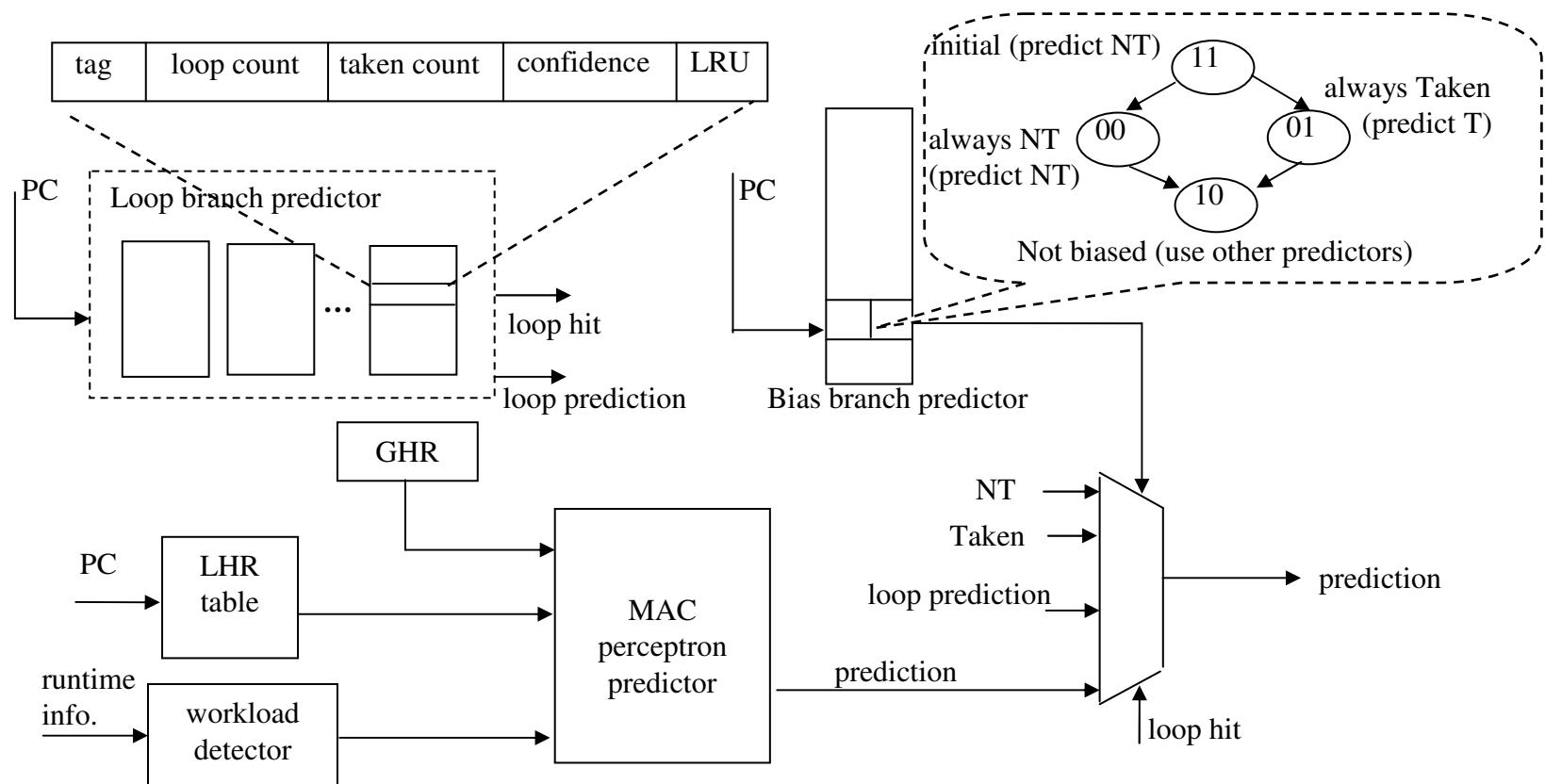


# Correlation-directed adaptation





# Overall scheme





# Summary

- Observations
  - Different workloads/branches need different information.
  - Perceptron weights  $\leftrightarrow$  Correlation
- Contributions
  - Profile-directed adaptation
  - Correlation-directed adaptation
  - Reducing aliasing from bias and loop branches
- Result
  - Significant improvement

**Thank you and Questions?**



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## References

- [1] I. K. Chen, J. T. Coffey, and T. N. Mudge, “Analysis of branch prediction via data compression”, *Proc. of the 7<sup>th</sup> Int. Conf. on Arch. Support for Programming Languages and Operating Systems* (ASPLOS-VII), 1996.
- [2] D. Jimenez and C. Lin, “Dynamic branch prediction with perceptrons”, *Proc. of the 7<sup>th</sup> Int. Symp. on High Perf. Comp. Arch* (HPCA-7), 2001.
- [3] D. Jimenez and C. Lin, “Neural methods for dynamic branch prediction”, *ACM Trans. on Computer Systems*, 2002.
- [4] S. MacFarling, “Combining branch predictors”, *Technical Report*, DEC, 1993.
- [5] A. Seznec, “Revisiting the perceptron predictor”, *Technical Report*, IRISA, 2004.
- [6] T.-Y. Yeh and Y. Patt, “Alternative implementations of two-level adaptive branch prediction”, *Proc. of the 22nd Int. Symp. on Comp. Arch* (ISCA-22), 1995.



# Predictor configuration

COMPONENT	CONFIGURATION	COST
Bias branch predictor	2293 entries	$2293 \times 2 = 4586$ bits.
Loop branch predictor	24 entries, 8-way set associative	1344 bits
Information	PC GHR: 100 bits LHR: 8 bits, 63 entries	$32 + 100 + 8 * 63 = 636$ bits
MAC perceptron predictor	<b>W0 table:</b> 61 entries, 8 bits each Other table sizes: 63, 55, 53, 53, 51, 49, 43, 41, 41, 39, 37, 37, and 35. <b>Total MAC entries:</b> 597 Each entry: 16 weights, 6 bits each <b>Control bits:</b> 2 bits each entry	$61 * 8 +$ $597 * 16 * 6 +$ $597 * 2$ $= 58994$ bits
Adaptation	Adaptation interval : $100000 * 2$ conditional branches	22 bits
Workload detection	Interval: 10000 (instructions / conditional branches)	82 bits
		Total : 65649 bits (less than $64 \times 1024 + 256 = 65792$ bits)