

## **CSE 240A Project 1 – Branch Predictor Contest**

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The predictor is based on the model of piecewise linear, which unifies the perceptron-based and path-based neural predictor.

The algorithm has two components: a prediction function and an update procedure. The following variables are used by the algorithm:

$W$ : a three-dimensional array of integers. The indices of this array are the branch address, the address of a branch in the path history, and the position in the history.  $W$  keeps track of correlations for every branch in the program.  $W[B,0,0]$  is the weight that keeps track of the tendency of branch  $B$  to be taken. This weight is the bias weight for  $B$ . Addition and subtraction on elements of  $W$  saturate at +127 and -128. The dimensions of the array are arbitrarily large, i.e., large enough to accommodate any access that might be made during the algorithm.

The indices of the  $W$  array must be limited to keep them from exceeding the practical bounds of an implementable branch predictor. I limit the first two indices by taking them modulo two integers  $n$  and  $m$ . In a realistic implementation,  $n$  and  $m$  would be chosen as powers of 2 to make the modulo operation a simple mask. We limit the third index by choosing an appropriate value  $h$  for the history length. Thus,  $W$  becomes an  $n*m*(h+1)$  3-dimensional array of 8-bit weights. For the

tuning of predictor, when  $n=m$ , it has better results.

*h*: The global history length. This is a small integer.

*GHR*: The global history register. This vector of bits accumulates the outcomes of branches as they are executed. Branch outcomes are shifted into the first position of the vector.

*GA*: An array of addresses. As branches are executed, their addresses are shifted into the first position of this array. Taken together, *GHR* and *GA* give the path history for the current branch to be predicted.

*output*: An integer. This integer is the value of the linear function computed to predict the current branch.

Overall, in my program, I choose  $h=15$ ,  $n=m=16$ . The optimal threshold is based on "Piecewise Linear Branch Prediction"<sup>[3]</sup>, which is  $2.14(h+1)+20.58$ . In my test cases, when threshold=33 the predictor has the optimal accuracy. The total size of tables I use is under the hardware budget, which is:

$$h+\log_2(m)*h+n*m*(h+1)*8=32843(\text{bits}) < 33024(32K+256)(\text{bits}).$$

I compare the piecewise linear predictor with Alpha 21264<sup>[4]</sup>, 2-level local prediction, global/local perceptron (GLP) and global perceptron (GP). From the following result it is clear to conclude that the piecewise linear predictor has the optimal average predict accuracy.

Trace	Alpha21264	2-level	GLP	GP	PWL
FP-1	3.357	4.147	2.023	2.454	3.071
FP-2	1.315	2.311	1.14	1.12	1.074
INT-1	8.587	14.08	6.27	7.451	6.244
INT-2	11.865	13.619	11.171	10.327	7.683
MM-1	8.918	10.813	7.7	7.68	7.434
MM-2	10.851	14.404	9.927	10.047	9.065
SERV-1	9.775	10.06	8.764	7.789	6.678
SERV-2	10.138	10.153	9.207	8.32	7.202
AVG	8.10075	9.948375	7.02525	6.8985	6.056375

#### References.

- [1] D. Jimenez, C. Lin, Dynamic Branch Predictor With Perceptrons, HPCA, 2001
- [2] D. Jimenez, C. Lin, Neural Methods for Dynamic Branch Prediction in ACM TOCS, 2002
- [3] D. Jimenez, Piecewise Linear Branch Prediction, ISCA, 2005
- [4] Kessler, The Alpha 21264 Microprocessor, IEEE Micro, 1999.