NC State University

Department of Electrical and Computer Engineering

ECE 463/563 (Prof. Rotenberg)

Project #2: Branch Prediction

REPORT TEMPLATE (Version 1.0)

by

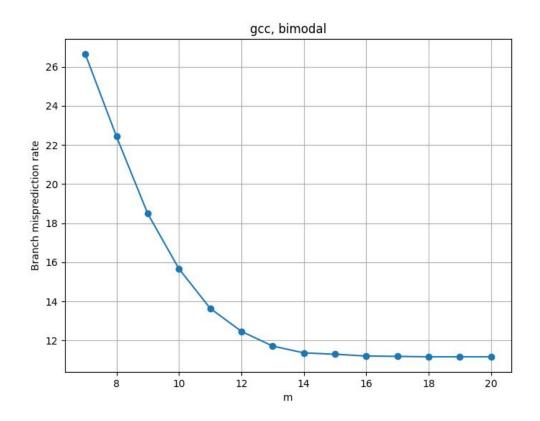
DEV DESAI

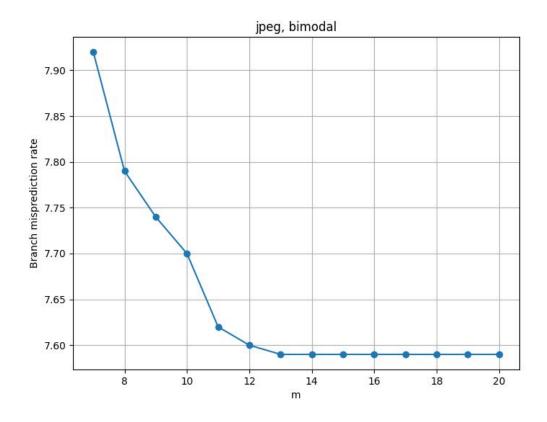
NCSU Honor Pledge: "I have neither given nor received unauthorized aid on this project."
Student's electronic signature: Dev Desai (sign by typing your name)
Course number: <u>563</u> (463 or 563 ?)

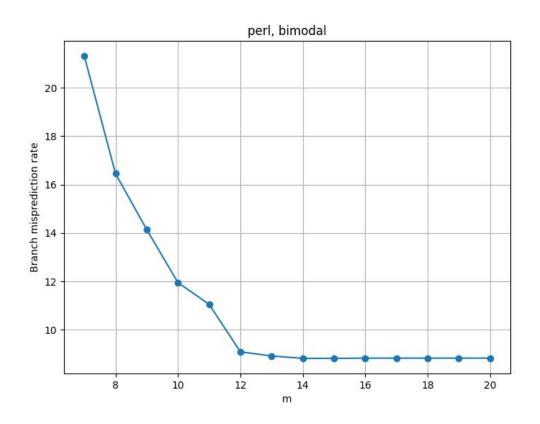
Grading Breakdown, Experiments, and Report

PART 1: BIMODAL PREDICTOR

- (a) [ECE463: 25 points] or [ECE563: 20 points] Gradescope will evaluate your simulator on the four validation runs "val_bimodal_1.txt", "val_bimodal_2.txt", "val_bimodal_3.txt", and "val_bimodal_4.txt", posted on the website for the BIMODAL PREDICTOR. Gradescope will also evaluate your simulator on one bimodal predictor mystery run. Each validation run and mystery run is worth ½ of the points for this part (5 or 4 points each). Gradescope must say that you match all four validation runs to get credit for the experiments with the bimodal predictor, however.
- (b) [ECE463: 25 points] or [ECE563: 20 points] Simulate BIMODAL PREDICTOR for different sizes $(7 \le m \le 20)$. Use the traces gcc, jpeg, and perl. [20 or 15 points] Graphs: Produce one graph for each benchmark. Graph title: "<benchmark>, bimodal". Y-axis: branch misprediction rate. X-axis: m. Per graph, there should be only one curve consisting of 14 datapoints (connect the datapoints with a line).







[5 points] Analysis:

- 1. [1 point] As the bimodal predictor's table size increases, the branch misprediction rate **Decreases**.
- 2. [2 points] For each benchmark, indicate the minimum value of m at which the misprediction rate bottoms-out (reaches its minimum) and indicate its minimum misprediction rate. Fill in the table below. (Use a precision of two digits after the decimal point for misprediction rate. That's what should be in your simulator's output file, in any case.)

Benchmark	Minimum "m" at which misprediction rate (two	Minimum misprediction rate
	digits of precision after the decimal point)	(two digits of precision after
	reaches its minimum	the decimal point)
gcc	m=18	11.17%
jpeg	m=13	7.59%
perl	m=14	8.82%

- 3. [1 point] At some point, increasing the bimodal predictor's table size is of no value. At this point, each static branch (*i.e.*, each static branch PC) is allocated a dedicated <u>entry</u> in the table. Given that interference among different static branches is eliminated at this point, the only way to improve accuracy further is a <u>better prediction algorithm</u>.
- 4. [1 point] I infer that **gcc** has **more** static branches (*i.e.*, unique static branch PCs) than **jpeg**, because **gcc** requires **more** table entries than **jpeg** before its misprediction rate bottoms-out.

PART 2: GSHARE PREDICTOR

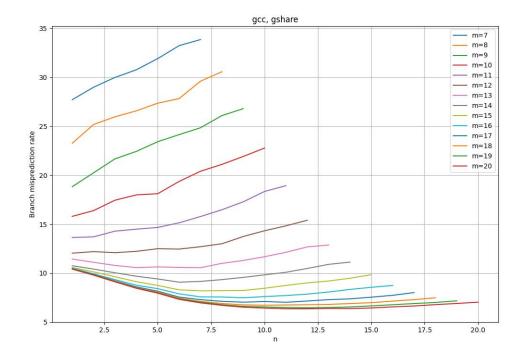
- (a) [ECE463: 25 points] or [ECE563: 20 points] Gradescope will evaluate your simulator on the four validation runs "val_gshare_1.txt", "val_gshare_2.txt", "val_gshare_3.txt", and "val_gshare_4.txt", posted on the website for the GSHARE PREDICTOR. Gradescope will also evaluate your simulator on one gshare predictor mystery run. Each validation run and mystery run is worth ½ of the points for this part (5 or 4 points each). Gradescope must say that you match all four validation runs to get credit for the experiments with the gshare predictor, however.
- (b) [ECE463: 25 points] or [ECE563: 20 points] Simulate GSHARE PREDICTOR for different sizes $(7 \le m \le 20)$, and for each size, *i.e.*, for each value of m, sweep the global history length n from 0 to m. Use only the trace gcc.
- [20 or 15 points] Graphs: Produce one graph for gcc. Graph title: "gcc, gshare". Y-axis: branch misprediction rate. X-axis: n (spanning n=0 to n=20). For this graph, there should be a total of 203 datapoints plotted as 14 curves. Datapoints having the same value of m (same predictor size) are connected with a line, i.e., one curve for each value of m. Note that not all curves have the same number of datapoints; see the listing below for the number of datapoints for each of the 14 curves, m=7 through m=20. The rationale for this graph is to study the effect of global history length for each predictor size.

```
m=7 curve has 8 datapoints: 0 \le n \le 7
m=8 curve has 9 datapoints: 0 \le n \le 8
m=9 curve has 10 datapoints: 0 \le n \le 9
m=10 curve has 11 datapoints: 0 \le n \le 10
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m=11 curve has 12 datapoints: $0 \le n \le 11$

m=12 curve has 13 datapoints: $0 \le n \le 12$

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m=13 curve has 14 datapoints: 0 \le n \le 13 m=14 curve has 15 datapoints: 0 \le n \le 14 m=15 curve has 16 datapoints: 0 \le n \le 15 m=16 curve has 17 datapoints: 0 \le n \le 16 m=17 curve has 18 datapoints: 0 \le n \le 17 m=18 curve has 19 datapoints: 0 \le n \le 18 m=19 curve has 20 datapoints: 0 \le n \le 19 m=20 curve has 21 datapoints: 0 \le n \le 20
```



[5 points] Analysis:

Insight: With the bimodal predictor (n=0: no global history), a given static branch is predicted using only a single 2-bit counter. With the addition of global history, that single counter is *specialized* or *multiplied* into many more counters. All these counters are used by the same static branch for more specialized predictions among its dynamic instances, promising higher accuracy *as long as there are adequate counters available in the table* (not just for this static branch, but for all static branches). Thus, with the addition of global history, there is a need for more 2-bit counters ... the key idea being that, at some point, bimodal cannot even take advantage of more counters (see your analysis section for bimodal, above) whereas gshare *can*. Summarizing: *Gshare needs an abundance of counters and, unlike bimodal, it can exploit abundant counters for higher accuracy*.

- 1. [0.5 points] At small table sizes, global history can <u>hurt</u> accuracy. This is because there are **too few** counters.
- 2. [0.5 points] At large table sizes, global history can <u>help</u> accuracy. This is because there are **abundant** counters.
- 3. [2.5 points] For each table size (m), indicate the **smallest** global history length (n) that yields the lowest misprediction rate (use a precision of two digits after the decimal point for misprediction rate; that's what should be in your simulator's output file,

in any case), indicate what that lowest misprediction rate is, and indicate the misprediction rate for bimodal. Fill in the table below.

m	Smallest global history length	Lowest	Bimodal misprediction
	(n) that yields the lowest	misprediction rate	rate (i.e., for $n=0$)
	misprediction rate (two digits of	(two digits of	(two digits of precision
	precision after the decimal	precision after the	after the decimal
	point)	decimal point)	point)
7	1	27.71%	26.65%
8	1	23.27%	22.43%
9	1	18.83%	18.49%
10	1	15.80%	15.67%
11	1	13.64%	13.65%
12	1	12.04%	12.47%
13	7	10.56%	11.72%
14	6	9.08%	11.37%
15	7	8.20%	11.30%
16	9	7.49%	11.21%
17	11	7.03%	11.19%
18	10	6.73%	11.17%
19	12	6.47%	11.17%
20	11	6.37%	11.17%

^{4. [0.5} points] The *smallest* bimodal predictor that achieves the best bimodal accuracy (lowest misprediction rate, **using two digits after the decimal point**, among all bimodal configurations) is as follows:

Best bimodal: m = 18, misp. rate = 11.17.

5. [0.5 points] The *smallest* gshare predictor that achieves the best gshare accuracy (lowest misprediction rate, **using two digits after the decimal point**, among all gshare configurations) is as follows:

Best gshare: m = 20, n = 11, misp. rate = 6.37%.

6. [0.5 points] In conclusion, with adequate predictor storage budget, gshare **rocks**.

PART 3: HYBRID PREDICTOR (ECE563 students only)

[ECE563: 20 points] Gradescope will evaluate your simulator on the two validation runs "val_hybrid_1.txt" and "val_hybrid_2.txt" posted on the website for the HYBRID PREDICTOR. Gradescope will also evaluate your simulator on two hybrid predictor mystery runs. Each validation run and mystery run is worth ¼ of the points for this part (5 points each).