Assignment — Evaluating Branch Prediction Schemes with an Out-of-Order Core in gem5

Goal: implement/compare several branch prediction schemes in gem5, run each with an out-of-order core on representative workloads, and analyze the effect of predictor design on IPC, misprediction rates, and overall processor performance.

1) Setup & baseline

- Build gem5 with the out-of-order CPU model (DerivO3CPU / O3CPU / O3-like CPU).
 - o Example configure/build:
 - scons build/X86/gem5.opt -j<nprocs> (adapt to ISA)
- 2. Choose workloads:
 - PARSEC / MiBench / simple compiled programs (e.g., gcc -O2 compile of small kernels, perl, fft, matrix multiply, bitmap, branch-heavy microbenchmarks)
 - Parsec Tutorial
 - MiBench Tutorial
 - o Include at least one **branch-heavy** and one **compute-heavy** workload.
- 3. Baseline run: run gem5 with the default branch predictor and OOO core to capture baseline stats.

2) Branch predictors to evaluate

Implement or enable and test the following predictors (at minimum):

- Bimodal (one-level) predictor (simple table, 2-bit saturating counters).
- **Gshare** (global history XOR with PC indexing).
- Local (per-PC) predictor (local history table + local counters).
- Tournament / hybrid predictor (combines global and local with selector).
- **Perceptron predictor** (if available/feasible) optional but high value.

If gem5 already provides some predictors (e.g., BiModeBP, GShareBP, TournamentBP, LocalBP, PerceptronBP), configure them. Otherwise implement missing ones in src/cpu/pred/ (or equivalent in the version you use).

Implementation notes:

- Predictor API: implement predict(), update() with branch PC, outcome, target.
- Keep parameters tunable: table size, counter bits, history length, perceptron weights length.
- Ensure predictor state can be instantiated from a command-line flag in gem5 (e.g., --bp-type=GShareBP).

3) OOO core configuration

 Use gem5's out-of-order pipeline (DerivO3CPU / O3CPU) with realistic ROB, IQ sizes. Example settings to vary:

ROB size: 128, 256IQ entries: 64, 128

- o Fetch width / issue width: 4
- Branch recovery penalty: use gem5 defaults, but record branchPredicted, branchMispredicted, and pipeline stalls.

Run all predictors on the **same** OOO configuration to isolate predictor effects.

4) Experiment methodology

- Fast-forward warmup: run with --cmd/checkpoints or run short input then enable stats collection for a "region of interest" (ROI).
- For each workload & predictor:
 - 1. Fast-forward to ROI (or use checkpoint).
 - 2. Run for a fixed number of committed instructions (e.g., 100M instructions) or simulate for a fixed simulated time.
 - 3. Collect gem5 stats (stats.txt).
- Repeat each configuration 3 times to check variability (optional but recommended).
- Keep all other microarchitectural parameters fixed across predictor runs.

5) Stats & metrics to collect

From gem5 stats.txt (or m5.stats), extract:

Primary metrics

- sim_seconds (sim time)
- system.cpu.<cpu>.committed instructions (or instrs committed)
- IPC = committed instructions / sim seconds or use gem5's IPC stat
- branchPredicted and branchMispredicted (or branchPredicted.??? depending on gem5 version)
- **Branch misprediction rate** = branchMispredicted / branchPredicted
- pipeline stalls (if available), fetch bubbles, squash count

Additional useful metrics

- I1d accesses, I1i accesses (see secondary effects)
- ROB occupancy, IQ utilization (if accessible)
- branchPred.MispredRecoveryCycles or stats for misprediction penalty (if present)
- Runtime breakdown (execute vs stall cycles)

6) Analysis tasks

- 1. For each workload and predictor, produce a table summarizing:
 - Predictor name & configuration (history length, table entries, counter bits)
 - o IPC
 - Branch prediction rate (accuracy and misprediction rate)
 - Average branch recovery penalty (or observed average stall cycles per misprediction)

2. Plot:

- Bar chart of IPC for each predictor (per workload)
- Line chart of misprediction rate vs predictor complexity (e.g., history length)
- Scatter plot: misprediction rate vs IPC
- 3. Explain:
 - Which predictor performed best overall and why.
 - o Where perceptron (if used) helps relative to tournament/gshare.
 - How predictor improvements translate (or not) into IPC improvements (discuss pipeline bottlenecks).
- 4. Statistical significance: comment on variability across runs.

7) Suggested command-line runs (example)

(adapt path/flags for your gem5 version and ISA; these are template-style)

Example: run DerivO3CPU with GShare predictor build/X86/gem5.opt configs/example/se.py \

- --cpu-type=DerivO3CPU \
- --caches --l2cache \
- --bp-type=GShareBP \
- --gshare-history=12 \
- --gshare-table-size=8192 \
- --cmd=/path/to/workload \
- --options="input args" \
- --maxinsts=100000000 \
- > out_gshare.log 2>&1

For Bimodal:

--bp-type=BiModeBP --bimodal-table-size=4096 --bimodal-counter-bits=2

For Tournament:

--bp-type=TournamentBP --global-history=12 --local-history=10 --chooser-size=4096

Perceptron (if available):

--bp-type=PerceptronBP --perceptron-h=32 --perceptron-table=4096

(If your gem5 doesn't accept these flags, extend the config script to accept them, or edit configs/ to create presets.)

8) Deliverables

- 1. **Code**: predictor implementations and config scripts (clean, commented).
- 2. **Run scripts**: shell/python scripts that reproduce the experiments.
- 3. **Data**: raw stats.txt for each run and a CSV summary.
- 4. Report (6–10 pages):
 - o Background & hypothesis
 - Experiment methodology (including warmup and ROI)
 - Results (tables + figures)
 - Analysis and discussion (why results look the way they do)
 - Conclusion and limitations
- 5. **Short presentation** (<=7 slides) summarizing key findings. We will use this for the viva.
- 6. This can be done in groups of two --- the report must clearly identify the exact contributions of the partners.
- 7. We will check for variance in the data that is generated by different groups. Please indicate the machine configuration of the machine on which the data is generated --- either from /proc or equivalent windows/mac places.

9) Grading rubric

- Correctness of implementations (25%) predictors implemented or configured correctly.
- Experiment design & reproducibility (20%) reproducible scripts, consistent methodology.
- Quality of analysis (25%) clarity of plots, correct interpretation of results, insight into why predictors perform as they do.
- Report & code quality (15%) readability, documentation, adherence to deliverable requirements.
- Viva (15)