Holding the line of performance and boosting optimizations for the PyTorch ecosystem.

Abstract

Goal: Quantifying PyTorch performance during the development loop.

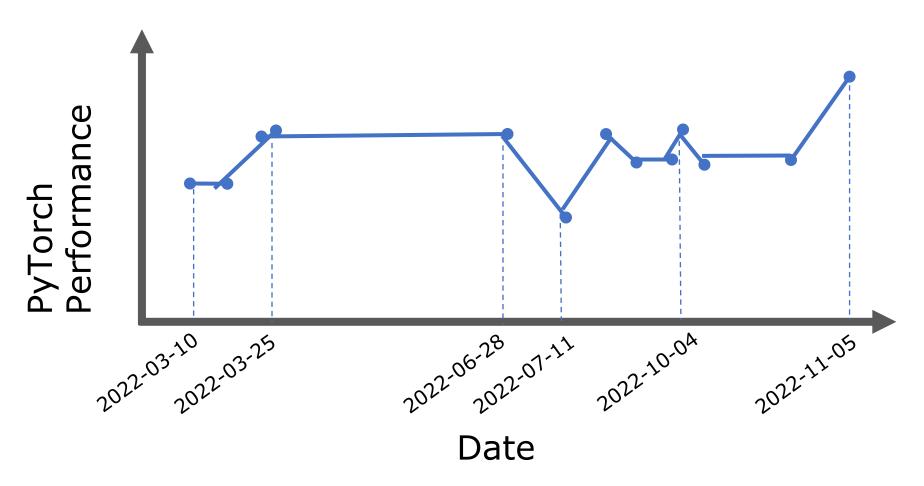


Fig. 1: Quantification of PyTorch performance in 2022.

Specifically, we answer the following questions:

- How do we know a GitHub PR speeds up or slows down PyTorch?
- How do we know an update in vendor libraries (e.g., CUDA) affect PyTorch performance?
- How do we find out the commit that caused PyTorch performance regression?

Challenges

Fast response Each test must finish in seconds instead of minutes or hours.

Coverage Need to cover both train and inference across a wide range of PyTorch models and devices.

Stableness Need to tune the benchmark environment to achieve stable results.

Technologies

Slicing model code We run one iteration per train/inference test with data prefetched to the device to achieve fast Time-To-Signal.

Tuning performance on public cloud We carefully tune the AWS/GCP hardware to stabilize performance results.

Bisection We implement a bisector to automatically find out which commit caused the performance signal, including both optimization and regression.

Performance Signals Detected

Date	Root Cause	Reason
2022-03-25	builder#1000	Speedup: cuDNN upgrade
2022-07-11	pytorch#65839	Regression: additional runtime check
2022-10-04	pytorch#84948	Speedup: enable new cuDNN front-end API

Boosting PyTorch backend development

Simple interface to add a new backend

1 # Example: torchscript backend

2 @create_backend

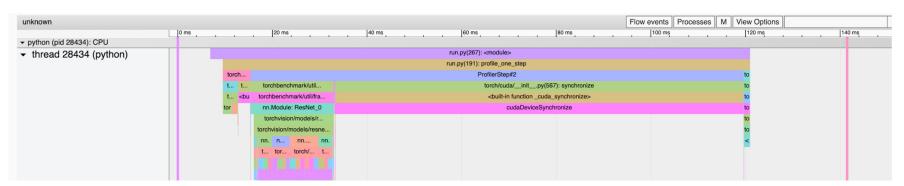
3 def torchscript(model: BenchmarkModel, args: List[str]):

module, example_input = model.get_module()

model.set_module(torch.jit.script(module, example_inputs))

Profile and run the backend

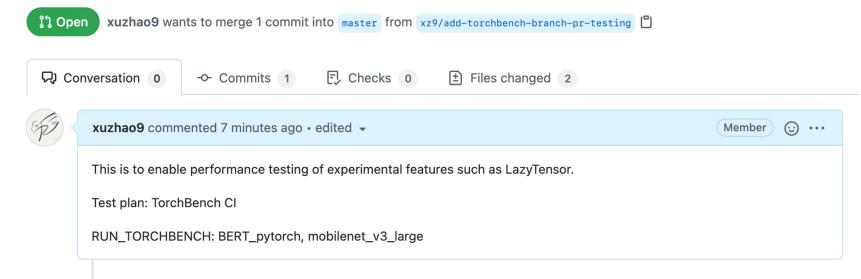
\$ python run.py resnet50 --backend torchscript --profile



Benchmarking-as-a-service

PR-level benchmarking With a magic word "RUN_TORCHBENCH" in PR, users can choose to benchmark their PR and visualize results at PyTorch HUD.

Enable the on-demand performance PR testing to run on a specified branch #64701



Metric	111fe61602	7bc72f5a2f	Speedup
nnc-dynamic:autogen-0	0.14310094044776633	0.14308511896524578	0.01%
nnc-dynamic:autogen-1	0.11164433404337615	0.11165716196410358	-0.01%
nnc-dynamic:autogen-10	0.017939746397314594	0.017773296852828933	0.94%
nnc-dynamic:autogen-11	0.02166838520206511	0.021501831093337385	0.77%
nnc-dynamic:autogen-12	0.12938609847333285	0.12939167249714956	-0.00%
nnc-dynamic:autogen-13	1.8119537853635848	1.8118110403884202	0.01%
nnc-dynamic:autogen-14	7.227453680243343	7.228049130644649	-0.01%

User customized benchmark Sometimes users want customized performance metrics or simply microbenchmarks, we support this through user-defined benchmark scripts. GPU memory leak and utilization signals are working-in-progress.

Nightly and Pre-release testing

TorchBench supports both Nightly and Prerelease testing CIs to catch performance bugs before production.

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