



HW-NAS-Bench: Hardware-aware Neural Architecture Search Benchmark



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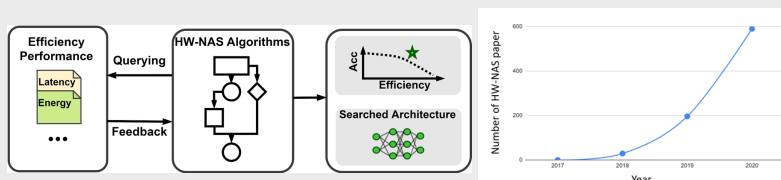
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ICLR'21 Spotlight

Background & Motivation

HW-NAS as an Automation Tool is on Growing Demand

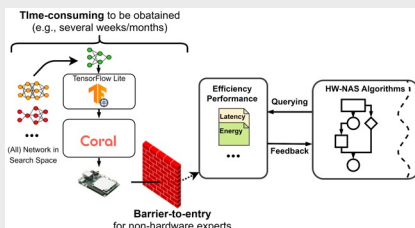
- HardWare-aware Neural Architecture Search (HW-NAS) **automatically** searches optimal architectures for **a target application and device**
- The number of HW-NAS research increases rapidly



Challenge 1: Non-trivial to Obtain Hardware-cost

HW-NAS requires **hardware-cost of (all) networks** in the search space

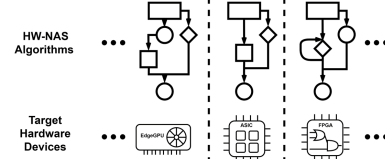
- Existing methods: hardware-cost look-up tables/device-specific estimator
- Limitations:
 - ⌚ **Time-consuming**
 - ⌚ **Barrier-to-entry** for non-hardware experts



Challenge 2: Difficult to Benchmark HW-NAS

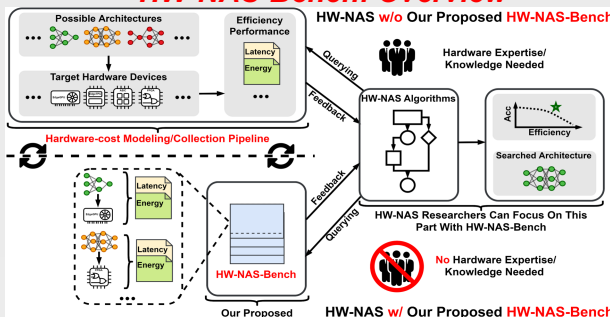
Difficult to benchmark different HW-NAS algorithms because of the **different adopted devices**

- Existing methods: Benchmarks focusing on accuracy/FLOPs/#Params/ server-level hardware-cost
- Limitations:
 - ⌚ **No/Limited** hardware-cost
 - ⌚ **FLOPs/#Params does not correlate well** with hardware-cost



The Proposed HW-NAS-Bench

HW-NAS-Bench: Overview



HW-NAS-Bench: Highlighted Features

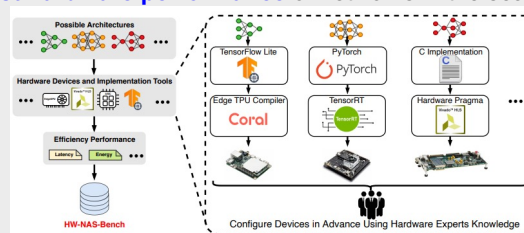
- **Six** devices, **three** categories
- **Two** search spaces: NAS-Bench-201 and FBNet search spaces
- **Both** energy and latency

Devices	Edge GPU	Raspi 4	Edge TPU	Pixel 3	ASIC-Eyeriss	FPGA
Collected Metrics	Latency (ms) Energy (mJ)	Latency (ms)	Latency (ms)	Latency (ms)	Latency (ms) Energy (mJ)	Latency (ms) Energy (mJ)
Collecting Method	Measured	Measured	Measured	Measured	Estimated	Estimated
Runtime Environment	TensorRT	TensorFlow Lite	Edge TPU Runtime	TensorFlow Lite	Accelerergy+TimeLoop / DNN-Chip Predictor	Vivado HLS
Customizing Hardware?	×	×	×	×	✓	✓
Category	Commercial Edge Devices			ASIC		FPGA

- 😊 **Democratize** HW-NAS research to **non-hardware experts**
 - Solved Challenge 1: Non-trivial to Obtain Hardware-cost
- 😊 **Facilitate a unified benchmark for HW-NAS**
 - Solved Challenge 2: Difficult to Benchmark HW-NAS

HW-NAS-Bench: Hardware-cost Collection Pipeline

- Implementation in each device is **optimized to make sure the best hardware performance** of networks in the search space



HW-NAS-Bench: Analysis

Analysis 1: Real Hardware-cost Is Necessary

Theoretical metrics (e.g., FLOPs/#Params) do **NOT** correlate well with real-measured/estimated hardware-cost

- Coefficient can be **as low as 0.36**

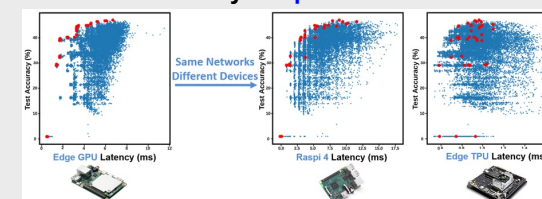
Analysis 2: Device-specific Cost Is Necessary

Hardware-cost on one device do **NOT** correlate well with hardware-cost on other devices

- Coefficient can be **as low as 0**

Analysis 3: Device-specific HW-NAS is Necessary

DNN architectures with the **optimal accuracy vs. hardware-cost trade-offs** in one device may **not** perform well in another device



HW-NAS-Bench: Easy-to-use APIs

- Create API

```
from hw_nas_bench_api import HWNASBenchAPI as HWAPI
hw_api = HWAPI("HW-NAS-Bench-v1.0.pickle", search_space="nasbench201")
```

- Get the real-measured/estimated hardware-cost

```
HW_metrics = hw_api.query_by_index(0, "cifar10")
```

- Example output:

Try it now:

```
==> Example to get use the hardware metrics
edgegpu_latency: 5.807418537139893 (ms)
edgegpu_energy: 24.226614330768584 (mJ)
raspi4_latency: 10.481976820010459 (ms)
edgegpu_latency: 0.9571811309997429 (ms)
pixel3_latency: 3.6058499999999998 (ms)
eyeriss_latency: 3.6456200000000001 (ms)
eyeriss_energy: 0.6872827644999999 (mJ)
fpga_latency: 2.57296 (ms)
fpga_energy: 18.01072 (mJ)
```



<https://github.com/GATECH-EIC/HW-NAS-Bench>