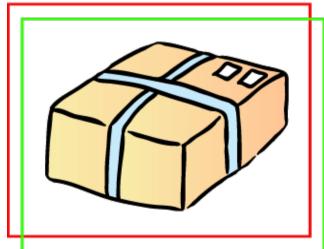


NNPACK



NNPACK

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NNPACK is an acceleration package for neural network computations. NNPACK aims to provide high-performance implementations of convnet layers for multi-core CPUs.

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NNPACK is not intended to be directly used by machine learning researchers; instead it provides low-level performance primitives to be leveraged by higher-level frameworks, such as Caffe, Torch, MXNet, Theano, Tensorflow, and Mocha.jl.

Requirements

Host system

- Linux or OS X host system
- x86-64 processor with AVX2 instruction set
 - NNPACK is optimized for Intel Skylake, but can run on Haswell & Broadwell processors too
 - o SSE2 instruction set can be targeted using --backend=psimd or --backend=scalar configuration options, but for performance reasons it is not recommended for production use
- ARMv7 processor with NEON instruction set
 - o VFP instruction set (including ARMv6 systems with VFPv2) can be targeted using --backend=scalar configuration option, but for performance reasons it is not recommended for production use.

Cross-compilation options:

- Android with x86/x86-64 (SSE2), ARMv7 with NEON, or ARM64 architecture
- WebAssembly for next-generation Web browsers
- Emscripten/Asm.js to run inside any modern Web browser
- Portable Native Client to run inside Google Chrome (no packaging required)
- Native Client (x86-64) to run as a packaged Google Chrome App

Features

- Fast convolution algorithms based on Fourier transform and Winograd transform.
 - Forward propagation performance on Intel Core i7 6700K vs BVLC Caffe master branch as of March 24, 2016 (protobufs from convnet-benchmarks, integration via caffe-nnpack):

Library	Caffe	NNPACK	NNPACK	NNPACK
Algorithm	im2col + sgemm	FFT-8x8	FFT-16x16	Winograd F(6x6, 3x3)
AlexNet:conv2	315 ms	129 ms	86 ms	N/A
AlexNet:conv3	182 ms	87 ms	44 ms	70 ms
AlexNet:conv4	264 ms	109 ms	56 ms	89 ms
AlexNet:conv5	177 ms	77 ms	40 ms	64 ms
VGG-A:conv1	255 ms	303 ms	260 ms	404 ms
VGG-A:conv2	902 ms	369 ms	267 ms	372 ms
VGG-A:conv3.1	566 ms	308 ms	185 ms	279 ms
VGG-A:conv3.2	1091 ms	517 ms	309 ms	463 ms
VGG-A:conv4.1	432 ms	228 ms	149 ms	188 ms
VGG-A:conv4.2	842 ms	402 ms	264 ms	329 ms
VGG-A:conv5	292 ms	141 ms	83 ms	114 ms
OverFeat:conv2	424 ms	158 ms	73 ms	N/A
OverFeat:conv3	250 ms	69 ms	74 ms	54 ms
OverFeat:conv4	927 ms	256 ms	272 ms	173 ms
OverFeat:conv5	1832 ms	466 ms	524 ms	315 ms

- Built-in expert-tuned kernels with very high performance:
 - o Fast Fourier transform
 - Winograd transform

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- Matrix-matrix multiplication (GEMM)
- o Matrix-vector multiplication (GEMV)
- o Max-pooling.
- Multi-threaded SIMD-aware implementations of neural network layers.
- Implemented in C99 and Python without external dependencies.
- Extensive unit tests using C++ and Google Test.
- Supports Native Client target and outperforms native Caffe/CPU when running inside Chrome.

Layers

- Convolutional layer
 - Training-optimized forward propagation (nnp_convolution_output)
 - Training-optimized backward input gradient update (nnp_convolution_input_gradient)
 - Training-optimized backward kernel gradient update (nnp_convolution_kernel_gradient)
 - Inference-optimized forward propagation (nnp_convolution_inference)
- Fully-connected layer
 - Training-optimized forward propagation (nnp_fully_connected_output)
 - Inference-optimized forward propagation (nnp_fully_connected_inference)
- Max pooling layer
 - Forward propagation, both for training and inference, (nnp_max_pooling_output)
- ReLU layer (with parametrized negative slope)
 - Forward propagation, both for training and inference, optionally in-place, (nnp_relu_output)
 - Backward input gradient update (nnp_relu_input_gradient)
- Softmax layer
 - Forward propagation, both for training and inference, optionally in-place (nnp_softmax_output)

Building

NNPACK can be build on OS X and Linux.

Install ninja build system

```
sudo apt-get install ninja-build || brew install ninja
```

Install PeachPy assembler and confu configuration system

```
[sudo] pip install --upgrade git+https://github.com/Maratyszcza/PeachPy [sudo] pip install --upgrade git+https://github.com/Maratyszcza/confu
```

Then clone NNPACK, install dependencies, configure, and build

```
git clone https://github.com/Maratyszcza/NNPACK.git
cd NNPACK
confu setup
python ./configure.py
ninja
```

Cross-compilation for Android

- Download and setup Android NDK
- Add ndk-build to PATH variable
- Navigate to NNPACK directory and setup dependencies (confu setup)
- Build NNPACK with ndk-build build system.

Cross-compilation for Emscripten/WebAssembly

Download and setup upstream version of Emscripten SDK

- Using emsdk, download, build and activate incoming version of Emscripten and Binaryen, and setup environment variables. \$EMSCRIPTEN should specify the path to activated Emscripten environment.
- Configure NNPACK with --target=wasm option.

Cross-compilation for Emscripten/Asm.js

- Download and setup Emscripten SDK
- Using emsdk, download, build and activate one of the environments, and setup environment variables. \$EMSCRIPTEN should specify the path to activated Emscripten environment.
- Configure NNPACK with --target=asmjs option.

Cross-compilation for Portable Native Client

- Download and setup Native Client SDK
- Set NACL_SDK_ROOT variable to a versioned SDK directory (e.g. /opt/nacl_sdk/pepper_49).
- Configure NNPACK with --target=pnacl option.

Cross-compilation for Native Client

- Download and setup Native Client SDK
- Set NACL_SDK_ROOT variable to a versioned SDK directory (e.g. /opt/nacl_sdk/pepper_49).
- Configure NNPACK with --target=x86_64-nacl-newlib (recommended) or --target=x86_64-nacl-gnu option.

Testing

NNPACK contains extensive test suite for transformation and neural network layers.

After configuration type ninja smoketest to run a set of quick tests, or ninja test to additionally NNPACK layers with parameters from AlexNet, VGG-A, and Overfeat-Fast networks (this will take a while).

Packaging

Binary packages need to distribute two files: include/nnpack.h and lib/libnnpack.a (also lib/libnnpack.so or lib/libnnpack.dylib if NNPACK was configured with shared library support).

Bindings

Deep Learning Frameworks

- Caffe2 natively supports NNPACK
- MXNet supports NNPACK for inference in convolutional layers, fully-connected, and max-pooling layers. See MXNet wiki for configuration instructions and performance benchmarks).
- szagoruyko/nnpack.torch integration of NNPACK into Lua Torch via ffi
- tiny-dnn header-only deep learning framework in C++11, which natively supports NNPACK.
- darknet-nnpack fork of Darknet framework with NNPACK support.
- Maratyszcza/caffe up-to-date integration of NNPACK (convolutional, fully-connected, max-pooling, and ReLU layers)
 into Caffe based on nnpack-pr branch in ajtulloch/caffe.
- Maratyszcza/caffe-nnpack older and unmaintained integration of NNPACK (convolutional layers only) into Caffe.
- See also discussion in Issue #1

Languages and Environments

- node-nnpack Node.js bindings
- peterhj/libnnpack Rust bindings

Users

- Facebook uses NNPACK in production.
- Prisma uses NNPACK in the mobile app.

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