

GPU TECHNOLOGY CONFERENCE

NVIDIA Performance Primitives & Video Codecs on GPU

Gold Room | Thursday 1st October 2009 | Anton Obukhov & Frank Jargstorff

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Overview

- Two presentations:
 - NPP (Frank Jargstorff)
 - Video Codes on NVIDIA GPUs (Anton Obukhov)
- NPP Overview
 - NPP Goals
 - How to use NPP?
 - What is in NPP?
 - Performance



What is NPP?

- C Library of functions (primitives) running on CUDA architecture
- API identical to IPP (Intel Integrated Performance Primitives)
- Speedups up to 32x over IPP
- Free distribution
 - binary packages for Windows and Linux (32- and 64 bit), Mac OS X
- Release Candidate 1.0: Available to Registered Developers now.
 - Final release in two weeks at http://www.nvidia.com/npp



NPP's Goals

- Ease of use
 - no knowledge of GPU architecture required
 - integrates well with existing projects
 - work well if added into existing projects
 - work well in conjunction with other libraries
- Runs on CUDA Architecture GPUs
- High Performance
 - relieve developers from optimization burden
- Algorithmic Building Blocks (Primitives)
 - recombine to solve wide range of problems



Ease of Use

- Implements Intel's IPP API verbatim
 - IPP widely used in high-performance software development
 - well designed API
- Uses CUDA "runtime API"
 - device memory is handled via simple C-style pointers
 - pointers in the NPP API are device pointers
 - but: host and device memory management left to user (for performance reasons)
- Pointer based API
 - pointers facilitate interoperability with existing code (C for CUDA) and libraries (cuFFT, cuBLAS, etc.)
 - imposes no "framework" on developers



Example

```
// allocate host source image
int hp;
Ipp8u * pHI = ippiMalloc 8u C1(w, h, &hp);
        // fill with some image content
testPattern 8u C1(pHI, hp, w, h);
        // allocated device source image
int sp;
Npp8u * pSI = nppiMalloc 8u C1(w, h, &sp);
        // copy test image up to device
cudaMemcpy2D(pSI, sp, pHI, hp, w, h,
             cudaMemcpyHostToDevice);
        // allocate device result image
int dp;
Npp8u * pDI = nppiMalloc 8u C1(w, h, &dp);
        // Filter mask and achor
NppiSize mask = \{5, 5\};
NppiPoint anchor = \{0, 0\};
NppiSize ROI = \{w - mask.width + 1,
                    h - mask.height + 1};
        // run box filter
nppiFilterBox 8u C1R(pSI, sp, pDI, dp,
                     ROI, mask, anchor);
```

What is in NPP?

- Only Image-Processing Functions
 - subset of "IPPI" library
 - ~300 functions
- Limited set of data-types supported
 - 8-bit per channel: 8u_C1, 8u_C4, 8u_AC4
 - high bit depth: 32s_C1, 32f_C1
- Conversion functions to and from most other IPPI formats



What is in NPP?

- Data exchange & initialization
 - Set, Convert, CopyConstBorder, Copy, Transpose, SwapChannels
- Arithmetic & Logical Ops
 - Add, Sub, Mul, Div, AbsDiff
- Threshold & Compare Ops
 - Threshold, Compare
- Color Conversion
 - RGB To YCbCr (& vice versa), ColorTwist, LUT_Linear
- JPEG
 - DCTQuantInv/Fwd, QuantizationTable

- Filter Functions
 - FilterBox, Row, Column, Max, Min, Dilate, Erode, SumWindowColumn/Row
- Geometry Transforms
 - Resize , Mirror, WarpAffine/Back/Quad, WarpPerspective/Back/Quad
- Statistics
 - Mean, StdDev, NormDiff, MinMax, Histogram, SqrIntegral, RectStdDev
- Computer Vision
 - Canny Edge Detector



Performance

- Relative performance compared to IPP
 - Measuring methodology?
- Scalability
 - Problem size
 - Number of processor cores
- Some aggregated numbers
 - Performance suite averages



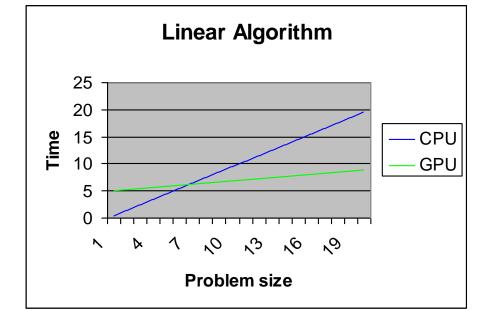
Performance Measuring Methodology

- Each primitive under test:
 - Is executed 25 times
 - Each iteration uses same data and same parameters
 - Data for GPU primitives is already on GPU (i.e. transfer times are not included in timings)
- All performance data gathered with single test application
 - test~2800 performance tests
 - most performance tests are simply repurposed functional tests
 - testing offset and oddly sized ROIs
 - testing various parameters
 - performance tests usually run at 720p and 2k x 2k image sizes



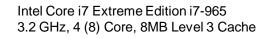
Scalability with Problem Size (1)

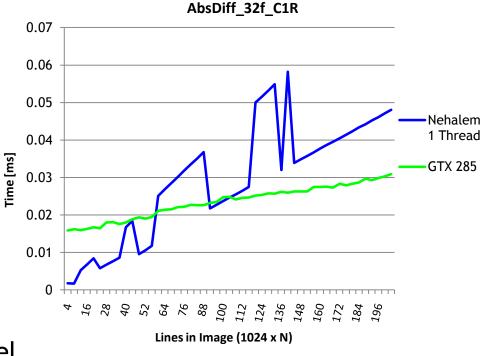
- Primtive: ippi/nppiAbsDiff_32f_C1R
 - computes per pixel absolute difference of two single-channel float image and stores result in third image
 - performance scales linearly with problem size
 - Time Plots: Lower is Better!
- Expected Results
 - linear with offsets O_{CPU} & O_{GPU} and slopes S_{CPU} & S_{GPU}
 - 0 < O_{CPU} < O_{GPU}
 - S_{CPU} > S_{GPU} > 0
- Where's the cross over point?



Scalability with Problem Size (2)

- Size from 1024x4 (16kB) to 1024x204 (~800kB)
- Offset & Slope:
 - CPU: O ~ 0 $\mu s, S$ ~25 $\mu s/100$ lines
 - GPU: O ~ 15 $\mu s,~S$ ~10 $\mu s/100~lines$
- Crossover:
 - CPU slow:
 - 48 lines = 48kPixel (4Byte) = 192kB
 - CPU fast:
 - 108 line = 108kPixel (4Byte) = 432kB
 - Compare: 720p: 1280 x 720 = 900kPixel

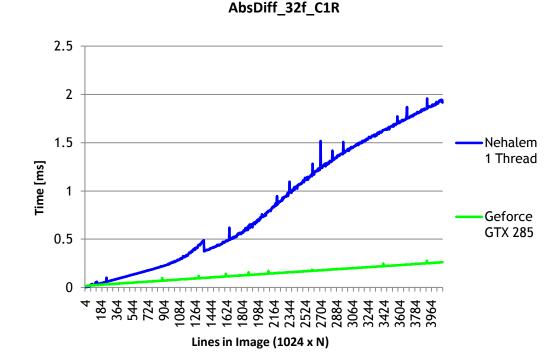






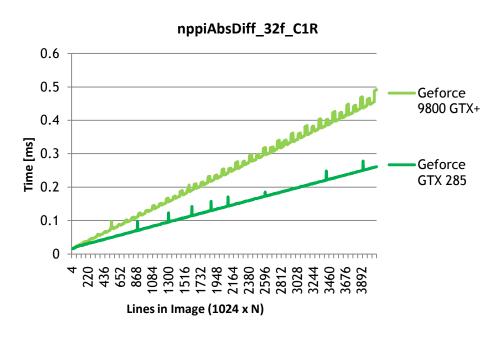
Scalability with Problem Size (3)

- Going in size up to 4096 lines
- GPU scales linearly
- Asymptotically CPU ~7.5x GPU
- CPU: Slope transition
 - Between ~1000 and ~3000 lines
 - 1000 lines = 4MByte image
 - 8MB level 3 cache size



Scalability With Number of Cores (1)

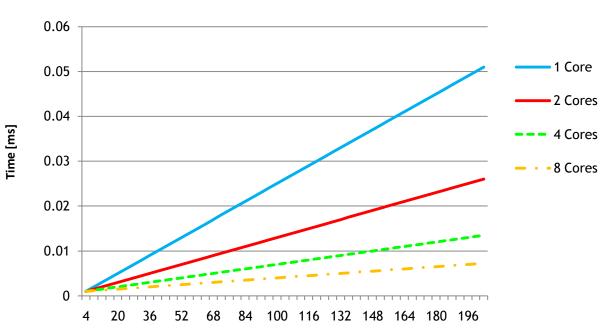
- For GPU not easy to control number of cores used.
- Compare two different GPUs/Graphics Cards:
 - Geforce 9800 GTX+: 16 SMs, 738MHz => 11808
 - Geforce GTX 285: 30 SMs, 648MHz => 19440
 - 19440/11808 = 1.64
- Chart at max size:
 - 9800 GTX: 480µs
 - GTX 285: 260µs
 - 4.8/2.6 = 1.84
- GPU scales linearly with number of SMs (cores) across full range of problem sizes.





Scalability With Number of Cores (2)

- Use ippSetNumThreads(int n); to control number of cores used.
- Expected Result:

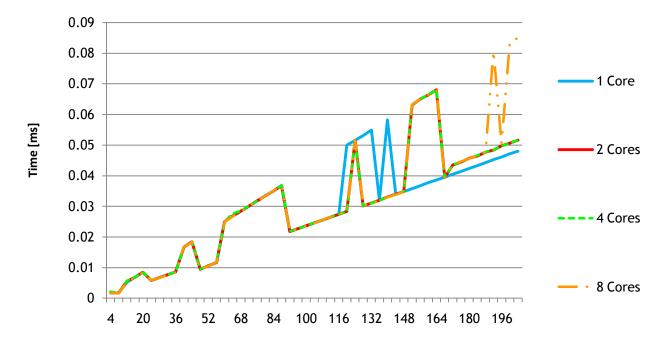


Lines in Image (1024 x N)



Scalability With Number of Cores (3)

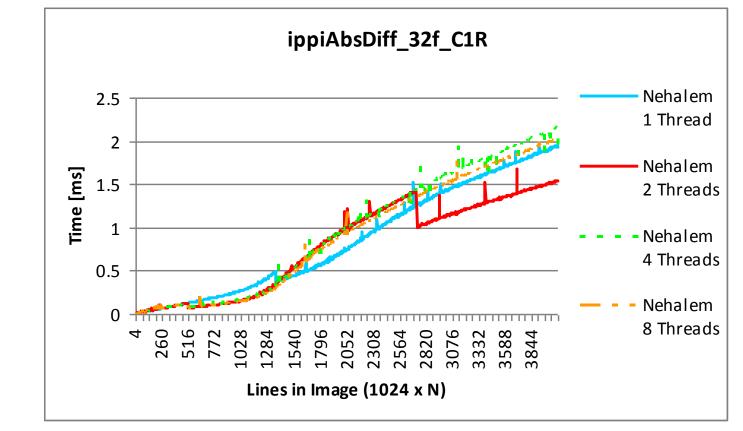
- CPU performance does not scale with number of cores, even for small problem sizes. ippiAbsDiff_32f_C1R: Intel Core i7 Extreme
- Actual Result:



Lines in Image (1024 x N)



Scalability With Number of Cores (4)



- Full range of image sizes on CPU
 - Not clear how

many threads are the best configuration for max performance.

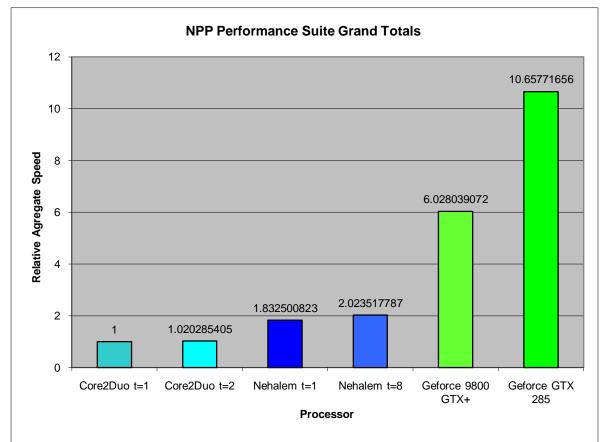
• CPU does not scale with number of Cores.

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Aggregate Performance Numbers (1)

- Average over 2800 performance tests.
 - each test gathers IPP and NPP processing times
 - performance tests are repurposed functional tests
 - run on 720p and 2k x 2k frames (mostly)





Aggregate Performance Numbers (2)

- Put into perspective:
 - NPP is 1.0 release
 - has been developed in 6 months
 - no processor specific optimizations*
 - all code compiled for compute 1.0 or 1.1
 - for the most part only optimized for memory coalescing
- Intel Core i7 vs. GTX 285
 - really different generations (GTX 285 uses 1.5 year old arch)
- That means there's still a lot of room for improvement.

* Exception: some statistics functions use atomics from compute capability 1.1.



Summary

- NPP
 - easy to integrate
 - provides substantial performance gains over highly optimized x86 code
 - 300 functions
- GPU/NPP Performance
 - scales extremely well with problem sizes and GPU type
 - room for performance improvements
 - particularly well suited for larger image sizes
- For questions regarding NPP please contact:
 - -<u>npp@nvidia.com</u>





GPU TECHNOLOGY CONFERENCE

Video Codecs on GPU

Fairmont Hotel, San Jose | 10.01.2009 | Anton Obukhov

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Motivation for the talk

Video encoding and decoding tasks require speedups as never before:

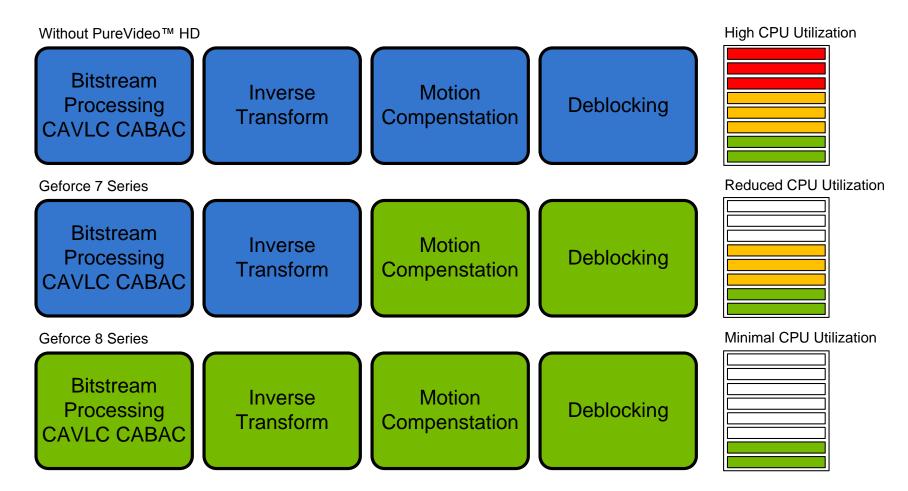




Encoding hi-res movie takes tens of hours on modern desktops Portable and mobile devices have unveiled processing power



Video capabilities evolution



Video encoding with NVIDIA GPU

Facilities:

- SW H.264 codec designed for CUDA
 - Baseline profile
 - Main profile
 - High profile

Interfaces:

- C library (NVCUVENC)
- Direct Show API
- Win7 MFT



Video decoding with NVIDIA GPU

Facilities:

- HW GPU acceleration of
 - H.264
 - VC1
 - MPEG2
- SW MPEG2 decoder designed for CUDA

Interfaces:

- C library (NVCUVID), HW & SW
- DXVA and Win7 MFT, HW only
- VDPAU library, HW only





Video processing with NVIDIA GPU

Facilities:

- SW pre- and post-processing library designed for CUDA
 - Noise Reduction
 - Deinterlacing 🖉
 - Polyphase Scaling
 - Color Processing
 - Deblocking
 - Detail enhance

Interfaces:

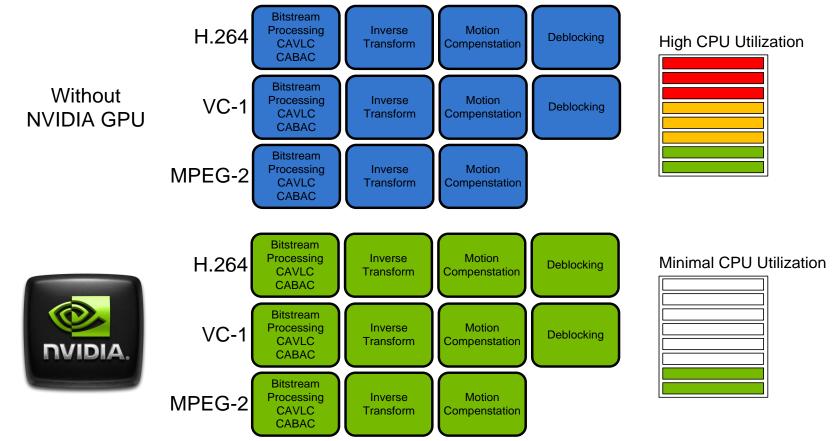
• VMR/EVR API





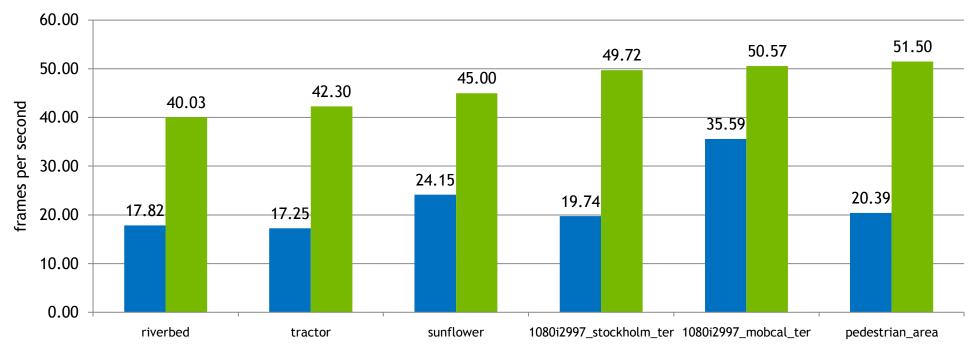
Benefits of Decoding with GPU

~100% Offload of 3 Major Video CODECs



Encode performance

CPU Encode GPU Encode



Frame size: 1080p Platform: 3.2 GHz quad core Nehalem, GeForce GTX 280 (240 core) GPU CPU encoder is x264 GPU encoder is NVIDIA H.264 CUDA encoder.

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Video encoding with NVIDIA GPU

Commercial applications for video transcoding with CUDA

- Badaboom
- Nero Move it
- CyberLink PowerDirector
- Loilo SuperLoiloscope
- Tons of them!



- What about 🙅 and 美?
- What about multi-GPU systems?



- What about 🙅 and 美?
 - Linux: only decoding acceleration with VDPAU
 - Mac OSX: QuickTime API

- What about multi-GPU systems?
 - NVIDIA H.264 encoder is going to support dual-GPU systems

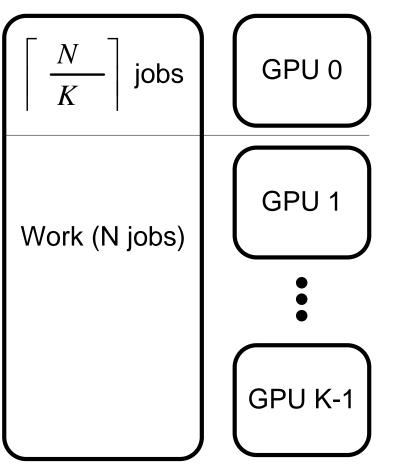


- Multi-GPU systems are commodity
- Programming for Multi-GPU systems is challenging





CUDA provides access to every GPU. How to make them all work efficiently?





There is a need for an open-source video codecs that can accelerate the transcoding pipeline using GPUs

Webinar 10/28/2009 9:00 AM - 11:00 AM PDT

- Multi-GPU techniques
- Application for video coding

https://www2.gotomeeting.com/register/628549827



Questions & Answers

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E-mail: aobukhov@nvidia.com

"Introducing a new Multi-GPU framework" webinar, 10/28/2009 9:00 AM - 11:00 AM PDT

https://www2.gotomeeting.com/register/628549827

