

OpenCL Imaging on The GPU: Optical Flow

James Fung Developer Technology, NVIDIA

Pyramidal Lucas Kanade Optical Flow in OpenCL

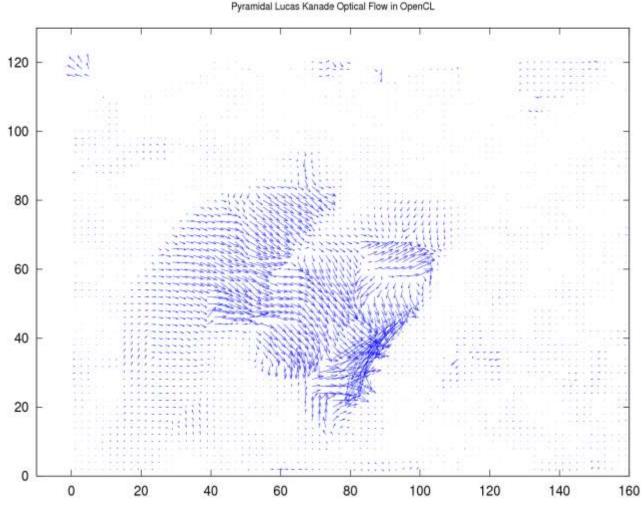
Algorithm by Jean-Yves Bouguet (Intel)
Implementation by James Fung (NVIDIA)

Optical Flow

- Calculate movement of selected points in pairs of images
- Applications:
 - Image stabilization
 - Feature tracking
 - Video encoding
- May be used to track
 - a few select points: sparse optical flow
 - All image points: dense optical flow
 - Computationally intensive!





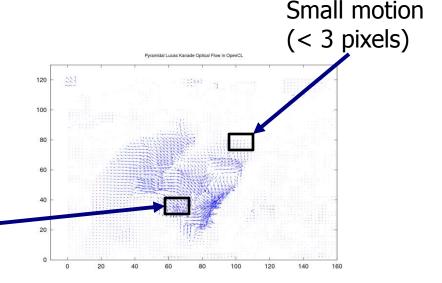


Algorithm

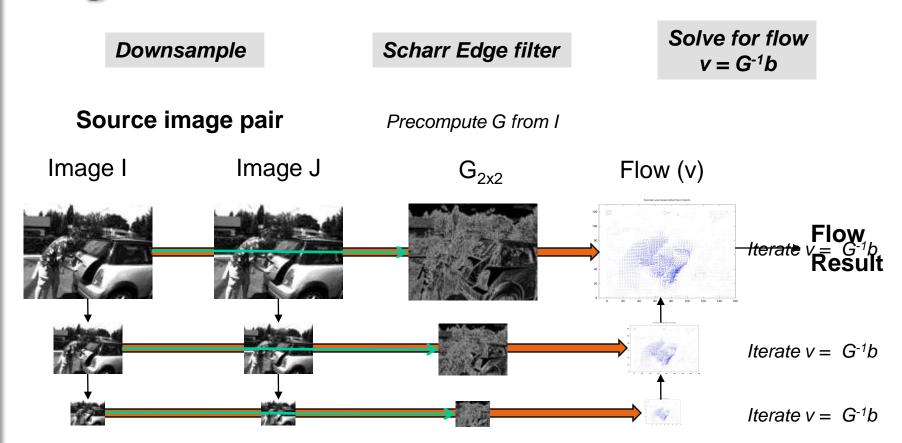
- For each point u on image I, find the corresponding point v on image J
- Calculate spatial gradient G in vicinity of each point u
- Iteratively solve for flow:

Calculate image mismatch vector \mathbf{b} Estimate motion $\mathbf{v} = \mathbf{G}^{1} \mathbf{b}(x',y')$ Update position $(x',y') = (x'+v_{x}, y'+v_{y})$ Repeat for N iterations or until convergence

Large Motion (3-7 pixels)



Algorithm

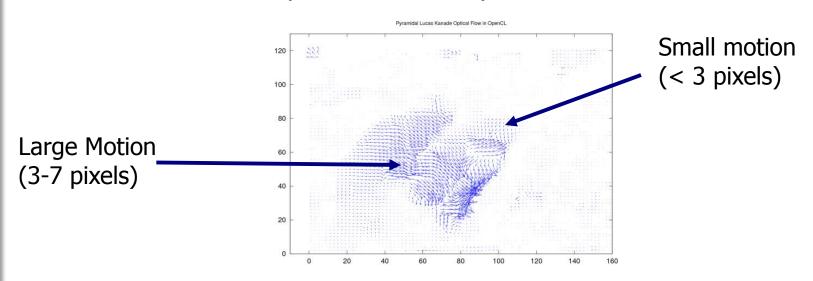


Optical Flow in OpenCL

- Embarrassingly parallel
- Computationally intensive
- Can take advantage of special-purpose GPU hardware:
 - Texture cache for 2D spatial locality
 - Hardware bilinear interpolation for sub-pixel lookups
 - Local Memory
- Can use OpenCL-OpenGL interoperability for visualization

Texture Cache

- Different areas of image have different amounts of motion, but motion is spatially coherent
 - Lookup window offset varies inside the image
 - Texture cache captures 2D locality



Hardware Interpolation

- Sub-pixel accuracy & sampling is crucial
- Between iterations, x,y is non-integer

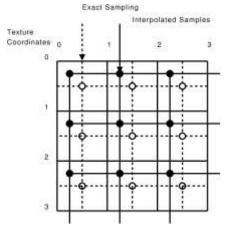
$$\partial I_{k}(x,y) = I^{L}(x,y) - J^{L}(x + g_{x}^{L} + v_{x}^{L}, y + g_{y}^{L} + v_{y}^{L})$$

$$b_{k} = \sum_{x=p_{x}-w_{x}}^{p_{x}+w_{x}} \sum_{y=p_{y}-w_{y}}^{p_{y}+w_{y}} \left[\frac{\partial I_{k}(x,y)I_{x}(x,y)}{\partial I_{k}(x,y)I_{y}(x,y)} \right]$$

Image level **L**Pixel center **p**Window **w**Guess **g** from previous level **L+1**

Use texture hardware linear interpolation

```
sampler_t bilinSample =
    ... | CLK_FILTER_LINEAR;
...
float Jsample = read_imagef( J_float,
    bilinSampler, Jidx+(float2)(i,j) ).x;
...
```

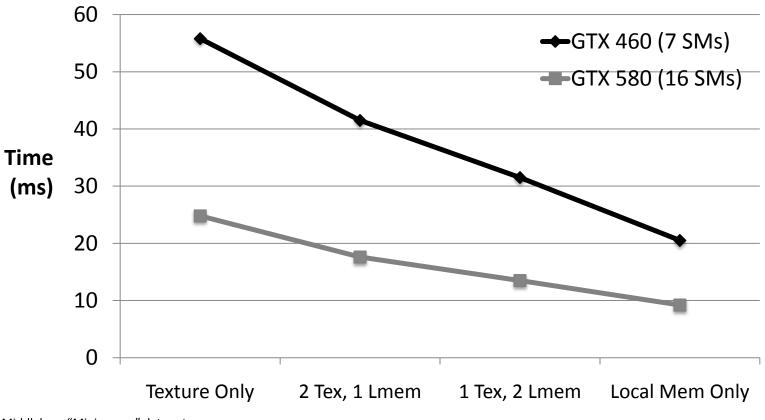


Using OpenCL Local Memory

- Lookup locations for I, Ix and Iy are static per iteration
- Can be explicitly cached in Local Memory
 - Local memory is on-chip
 - Local memory is faster to access, useful for repeated access

K H RON OS

Effect of moving from Texture to Local Memory Usage



Middlebury "Minicooper" data set

Frames: 10-11 Resolution: 640 x 480 Greyscale

Convergence: 0.0004px
Flow compute time for 3 pyramid levels shown

Local/Texture Buffers

Optical Flow Demo

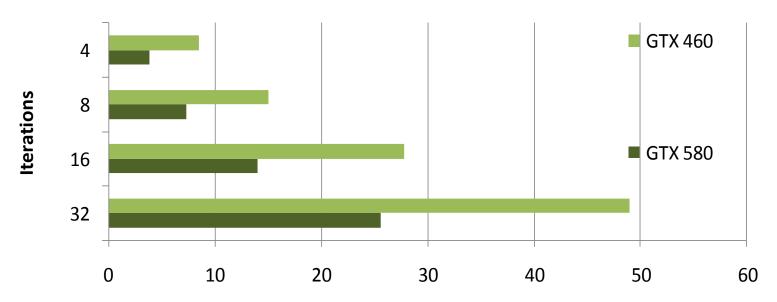
- Pyramidal Lucas Kanade
 Optical Flow
- Visualization done on GPU by sharing data between OpenGL & OpenCL



K H RON OS

Optical Flow Performance

LK Pyramidal Optical Flow GTX 460 & GTX 580



Time (ms) (lower is better)

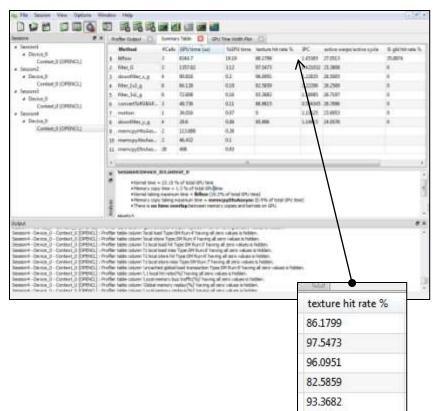
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Convergence: 0.0004px

Pyramid 3-level flow compute time shown

OpenCL Performance Analysis with the CUDA Visual Profiler 4.0

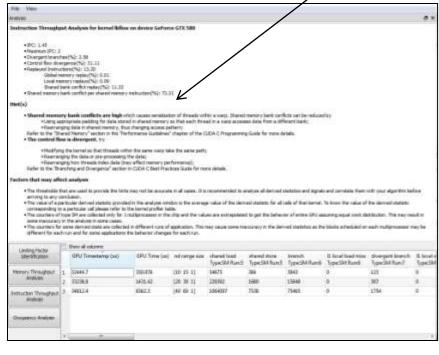


Shared bank conflict replay(%): 11.33

• Shared memory bank conflict per shared memory instruction(%): 72.21

t(s)

Shared memory bank conflicts are high which causes serialization of
 Using appropriate padding for data stored in shared memory so the
 Rearranging data in shared memory, thus chapping access pattern





Thank You!

Contact:

James Fung jfung@nvidia.com