



# Lecture 13

## Image filters

## OpenCL

## GPU computing with GLSL

## OpenGL Compute shaders



## Lecture questions

- 1) What kind of devices will OpenCL run on?**
- 2) What does an OpenCL work group correspond to in CUDA?**
- 3) What geometry is typically used for shader-based GPU computing?**
- 4) Are scatter or gather operations preferable? Why?**



## Lab 5

- **Image filtering with shared memory**
  - **Low-pass filters**
    - **Median filter**

**New lab last year. Worked well as continuation of previous image filtering lab.**



## Lab 6

- OpenCL
- Reduction
- Sorting using bitonic sort

**Also new last year, with good results.**



## Image filters (lab 3 and 5)





## Linear filters: Convolution

### Box filter

1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1

/25

### Gaussian approximation

1	4	6	4	1
4	16	24	16	4
6	24	36	24	6
4	16	24	16	4
1	4	6	4	1

/256

### And others

1	0	1
2	0	2
1	0	1

0	-1	0
-1	4	-1
0	-1	0



# Information Coding / Computer Graphics, ISY, LiTH

## Separable filters

$$\begin{matrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{matrix} \text{/5} \oplus \begin{matrix} 1 & 1 & 1 & 1 & 1 \end{matrix} /5 = \begin{matrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{matrix} \text{/25}$$

$$\begin{matrix} 1 \\ 4 \\ 6 \\ 4 \\ 1 \end{matrix} \text{/16} \oplus \begin{matrix} 1 & 4 & 6 & 4 & 1 \end{matrix} /16 = \begin{matrix} 1 & 4 & 6 & 4 & 1 \\ 4 & 16 & 24 & 16 & 4 \\ 6 & 24 & 36 & 24 & 6 \\ 4 & 16 & 24 & 16 & 4 \\ 1 & 4 & 6 & 4 & 1 \end{matrix} \text{/256}$$



## Repeated box filters converge to Gaussian!

$$\begin{matrix} 1 \\ 4 \\ 6 \\ 4 \\ 1 \end{matrix} /16 = \begin{matrix} 1 \\ 2 \\ 1 \end{matrix} /4 \oplus \begin{matrix} 1 \\ 2 \\ 1 \end{matrix} /4 = \begin{matrix} 1 \\ 1 \end{matrix} /2 \oplus \begin{matrix} 1 \\ 1 \end{matrix} /2 \oplus \begin{matrix} 1 \\ 1 \end{matrix} /2 \oplus \begin{matrix} 1 \\ 1 \end{matrix} /2$$



Central limit theorem  
Compare to dice!



## **Non-linear filters**

**Median filter**

**Outputs median of neighborhood.**

**Requires some method to find the median.**

**Important application noise suppression.  
Preserves edges!**

**Separable only as approximation.**



## Median filters





## How to filter edges

The filter kernel reaches outside the image!

My answer: clamp!

Solved for you in the Lab 5 code.

Why? Avoid branching!

```
if (x < imagesizex && y < imagesizey)
{
    // Filter kernel (simple box filter)
    sumx=0;sumy=0;sumz=0;
    for(dy=-kernelsizey;dy<=kernelsizey;dy++)
        for(dx=-kernelsizex;dx<=kernelsizex;dx++)
    {
        // Use max and min to avoid branching!
        int yy = min(max(y+dy, 0), imagesizey-1);
        int xx = min(max(x+dx, 0), imagesizex-1);

        sumx += image[((yy)*imagesizex+(xx))*3+0];
        sumy += image[((yy)*imagesizex+(xx))*3+1];
        sumz += image[((yy)*imagesizex+(xx))*3+2];
    }
    out[(y*imagesizex+x)*3+0] = sumx/divby;
    out[(y*imagesizex+x)*3+1] = sumy/divby;
    out[(y*imagesizex+x)*3+2] = sumz/divby;
}
```



## In the lab

1. Shared memory

Use shared memory to reduce global memory access.  
Major part of the lab!

2. Separable filters

Easy if step 1 is done right.

3. Weighted kernels

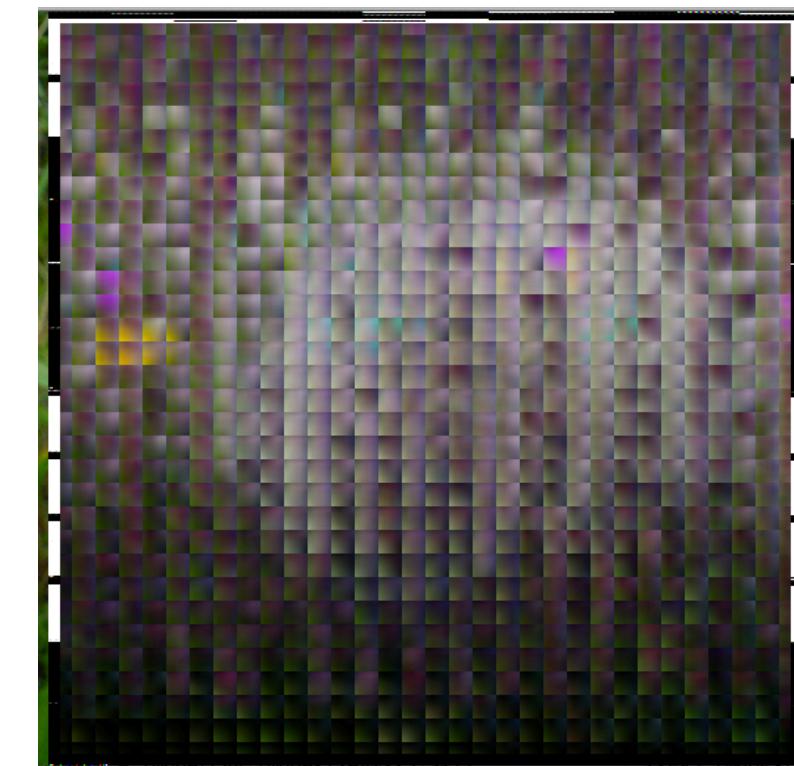
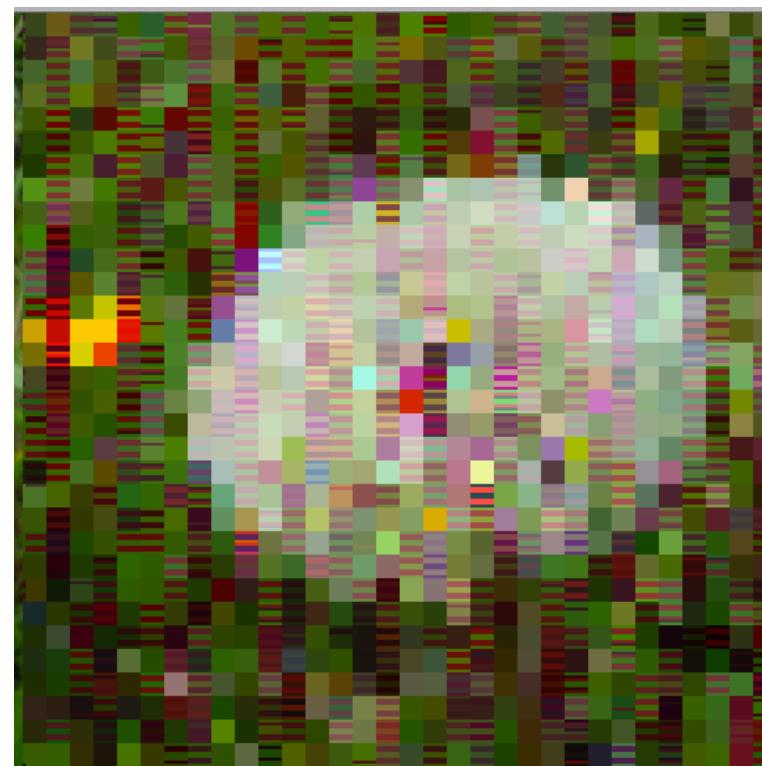
One size is enough.

4. Median filter

Variable size, modest demands.



## Bonus: Unintentional fun!



**Coding filters in CUDA is like a box of chocolate...**