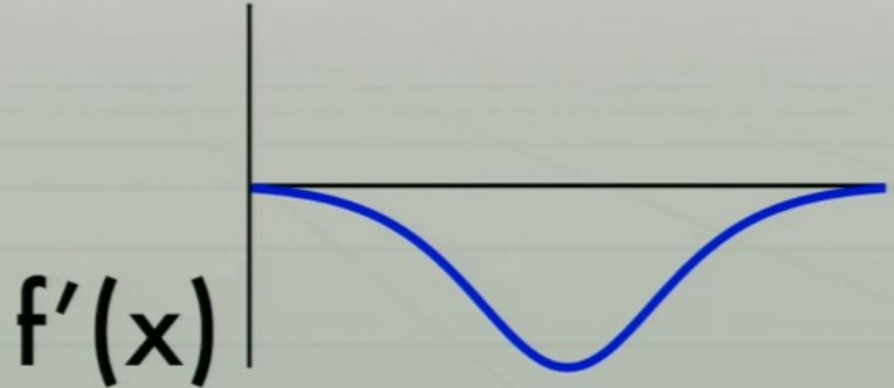
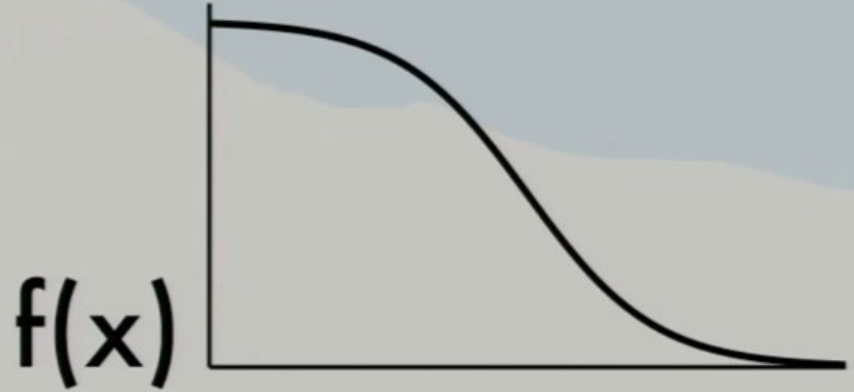


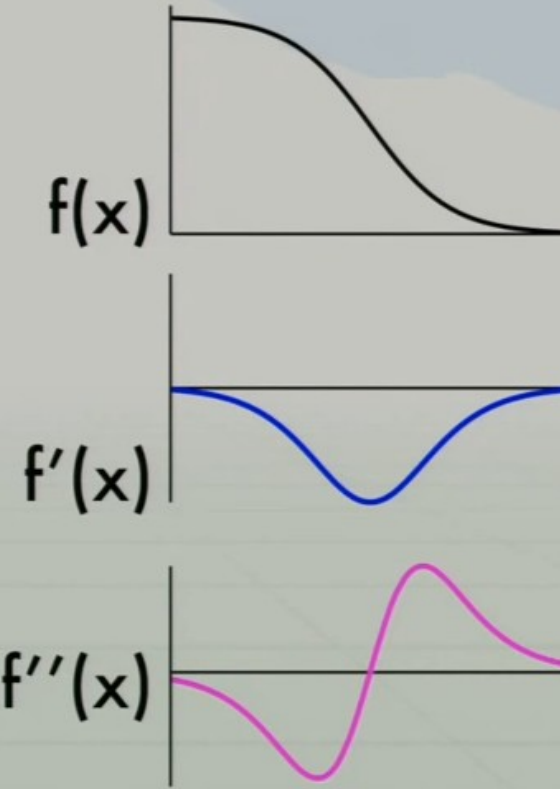
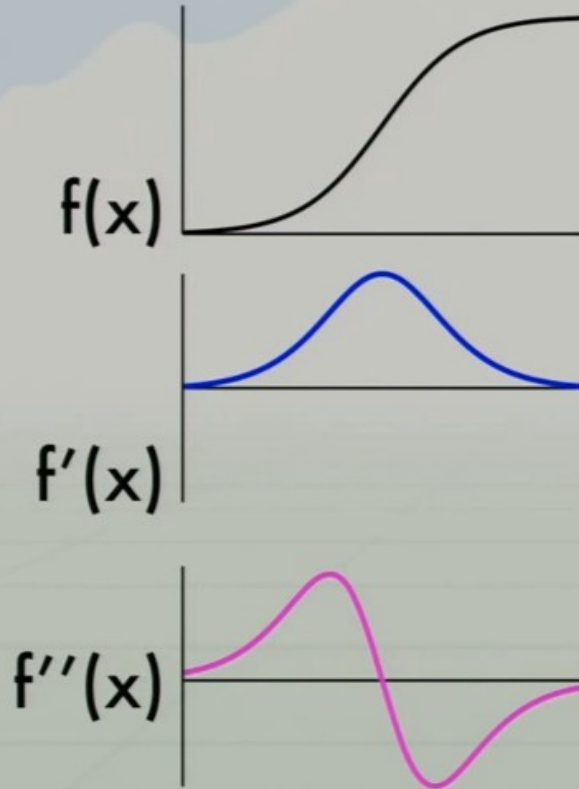
Finding edges

- Could take derivative
- Find high responses
- Sobel filters!
- But...
- Edges go both ways
- Want to find extrema



2nd derivative!

- Crosses zero at extrema



Laplacian (2nd derivative)!

- Crosses zero at extrema

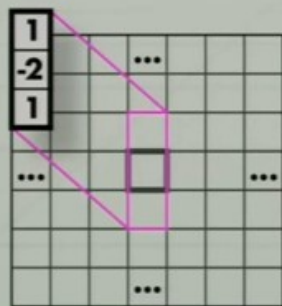
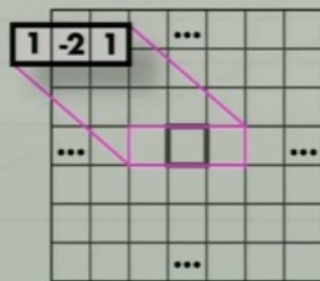
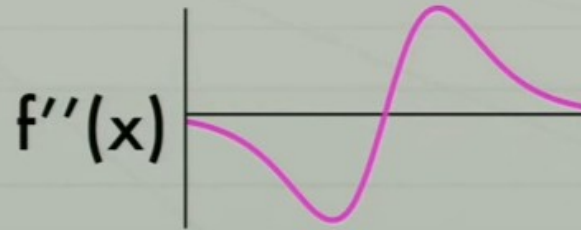
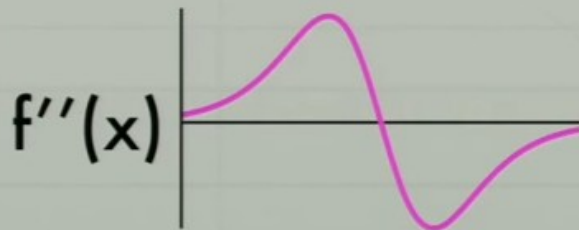
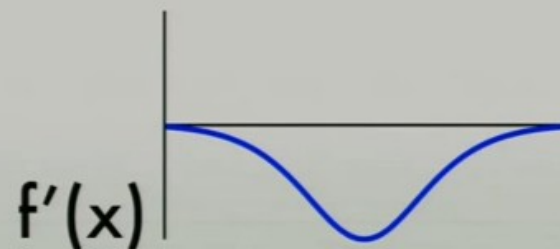
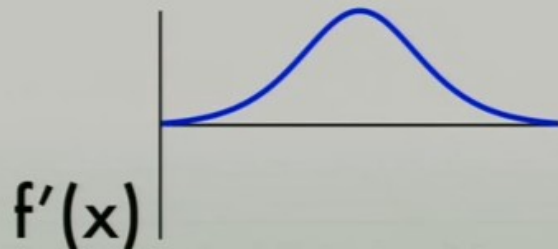
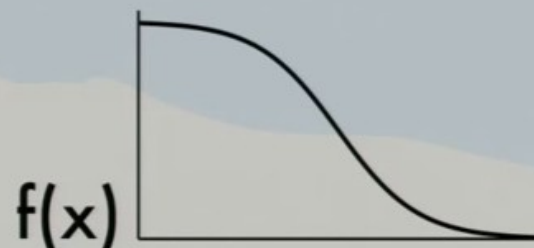
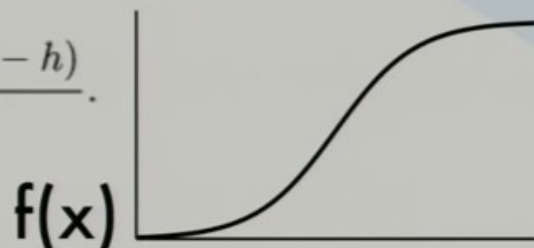
- Recall:

$$- f''(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - 2f(x) + f(x-h)}{h^2}.$$

- Laplacian:

$$- \Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

- Again, have to estimate $f''(x)$:



Laplacians

- Laplacian:

$$- \Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$


$$\left(\begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} * \begin{array}{|c|} \hline \text{Grid} \\ \hline \end{array} \right) + \left(\begin{bmatrix} 1 & -2 & 1 \end{bmatrix} * \begin{array}{|c|} \hline \text{Grid} \\ \hline \end{array} \right)$$

Laplacians

- Laplacian:

$$\Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

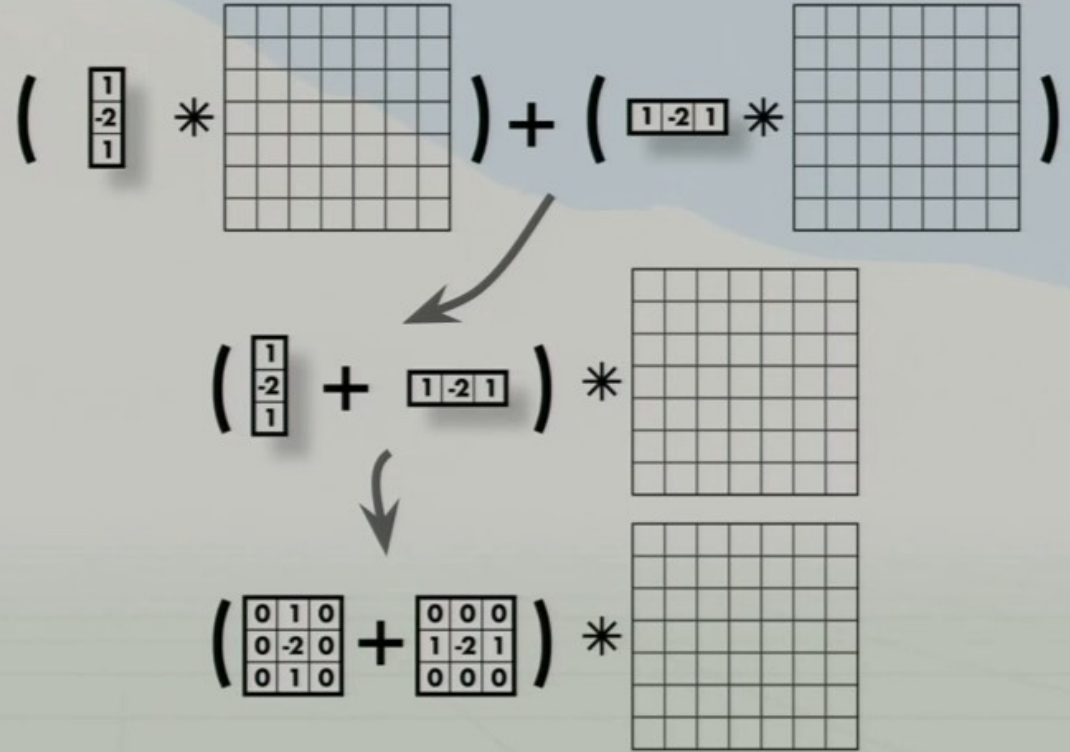
$$\left(\begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} * \begin{array}{|c|c|c|c|c|c|} \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline \end{array} \right) + \left(\begin{bmatrix} 1 & -2 & 1 \end{bmatrix} * \begin{array}{|c|c|c|c|c|c|} \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline \end{array} \right)$$

$$\left(\begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 & -2 & 1 \end{bmatrix} \right) * \begin{array}{|c|c|c|c|c|c|} \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline & & & & & \\ \hline \end{array}$$


Laplacians

- Laplacian:

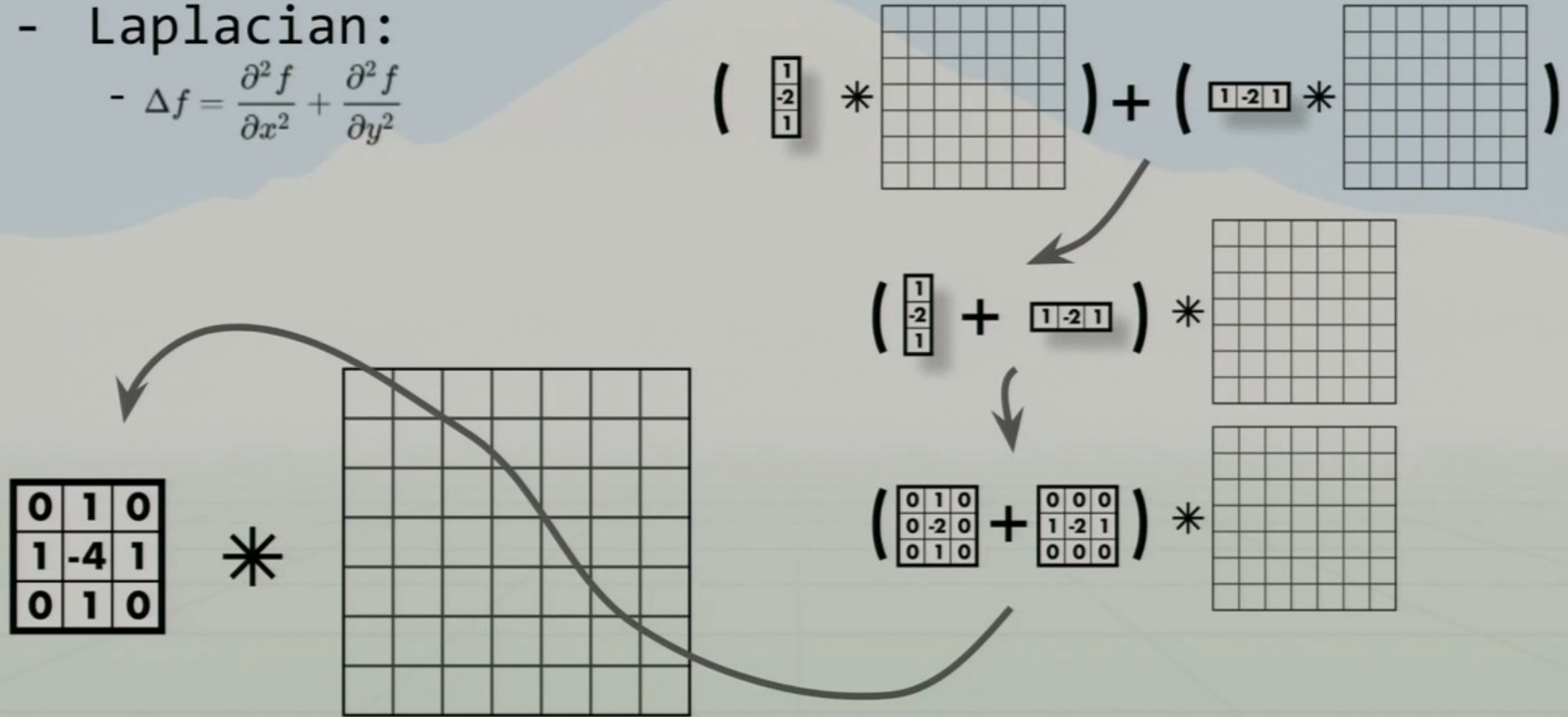
$$\Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

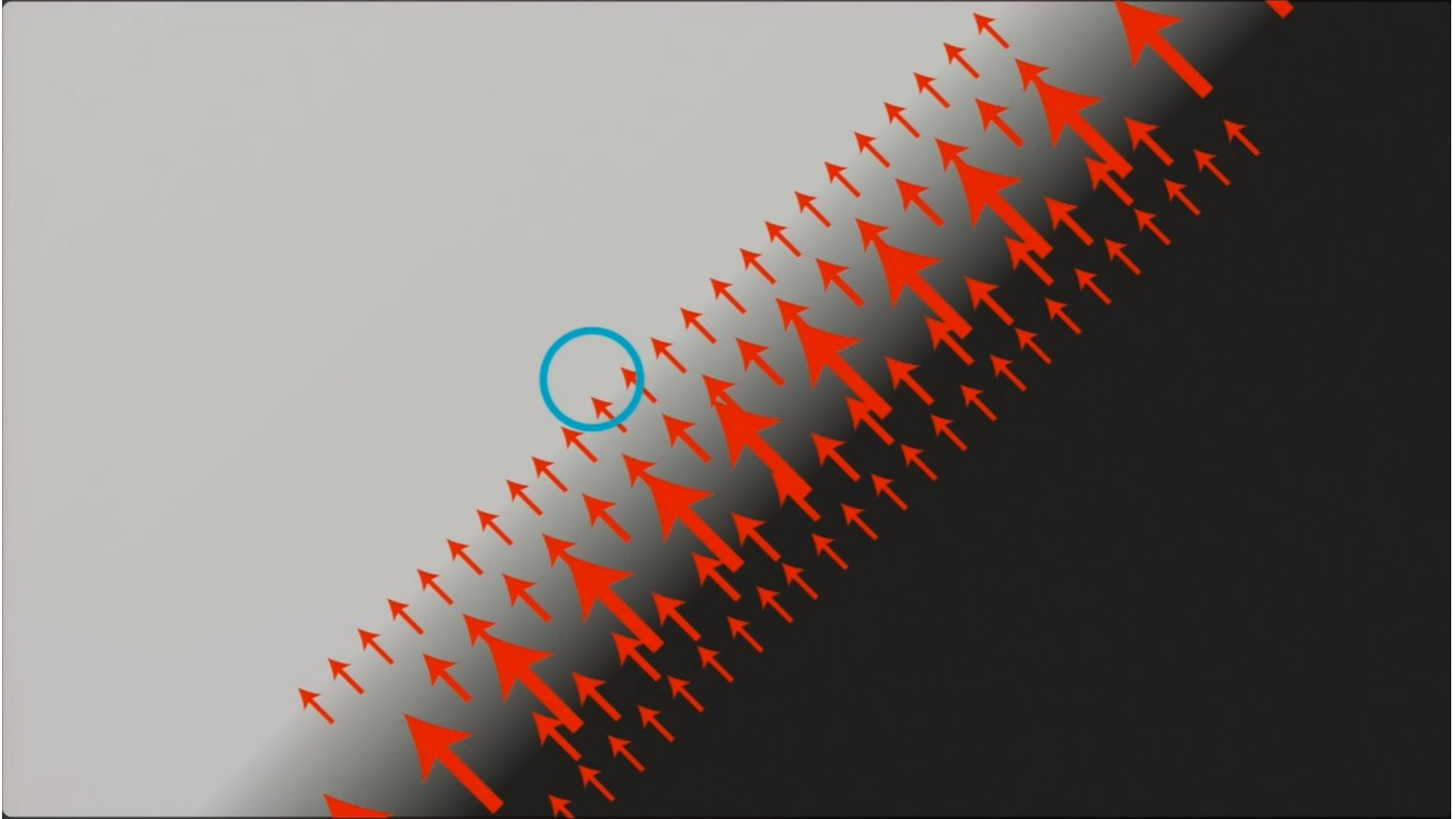


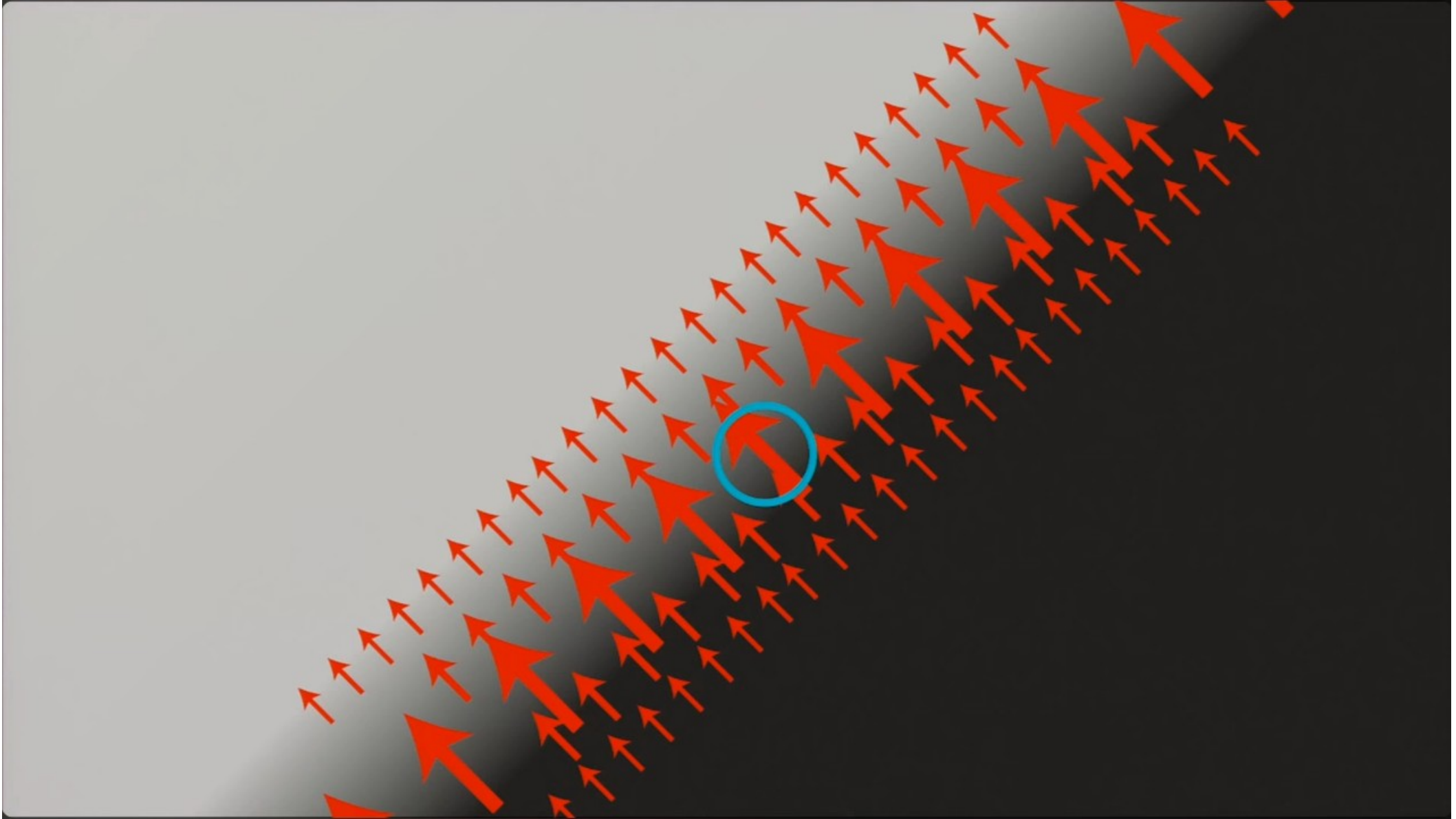
Laplacians

- Laplacian:

$$\Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$







Laplacians

- Laplacian:

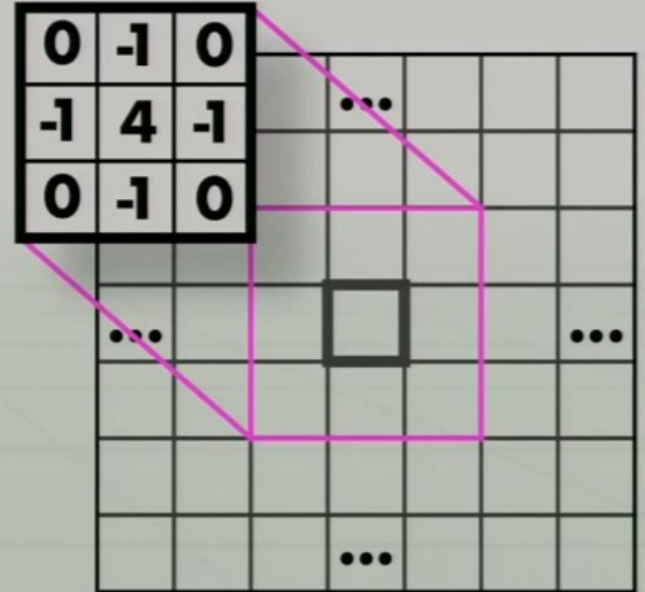
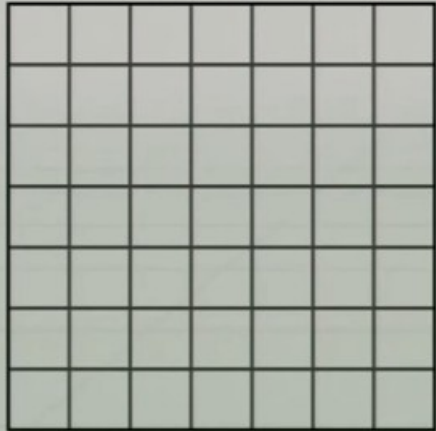
$$\Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

- Negative Laplacian, -4 in middle

- Positive Laplacian --->

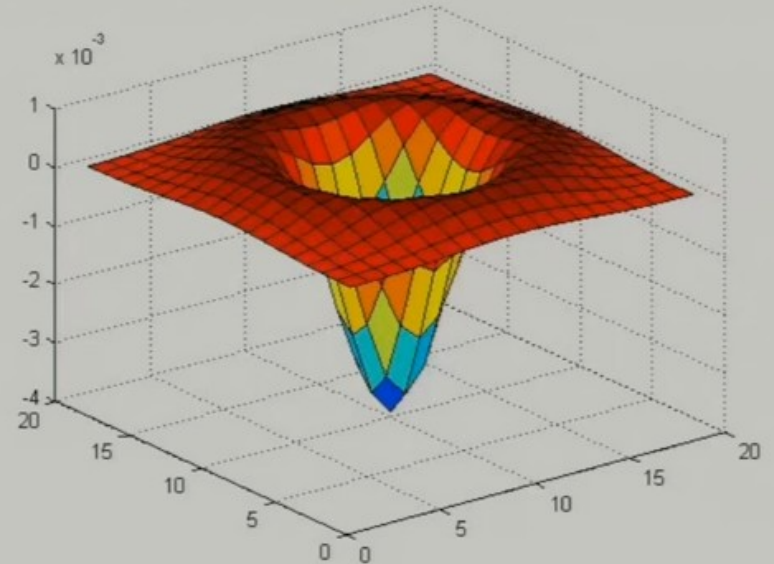
0	1	0
1	-4	1
0	1	0

*



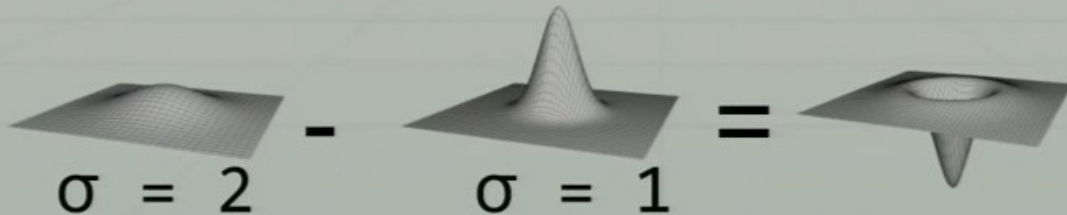
Laplacians also sensitive to noise

- Again, use gaussian smoothing
- Can just use one kernel since convs commute
- Laplacian of Gaussian, LoG
- Can get good approx. with 5x5 - 9x9 kernels



Difference of Gaussian (DoG)

- Gaussian is a low pass filter
- Strongly reduce components with frequency $f < \sigma$
- $(g * I)$ low frequency components
- $I - (g * I)$ high frequency components
- $g(\sigma_1) * I - g(\sigma_2) * I$
 - Components in between these frequencies
- $g(\sigma_1) * I - g(\sigma_2) * I = [g(\sigma_1) - g(\sigma_2)] * I$



DoGs

