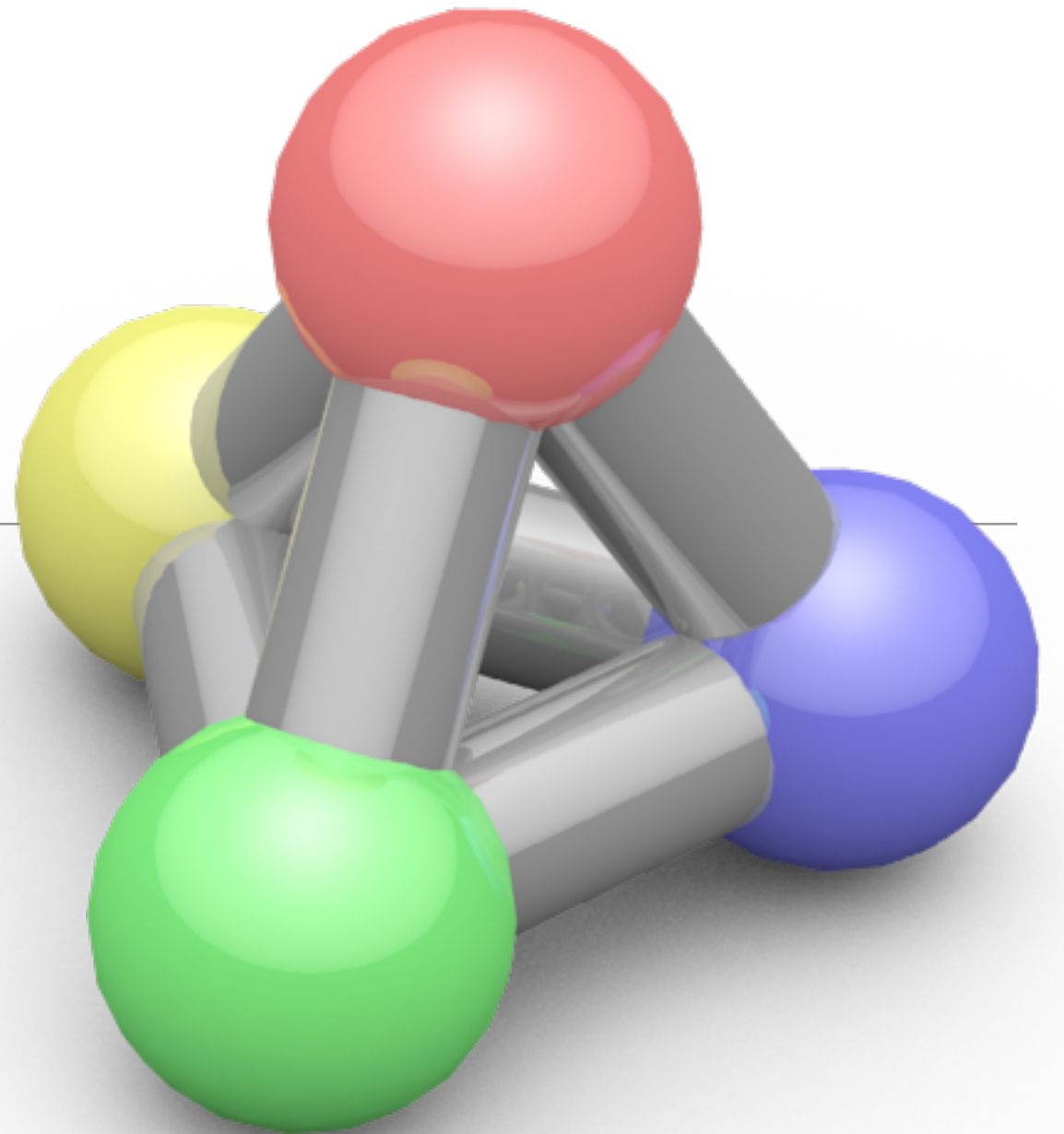


# Digital Image Effects

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CPSC 453 – Fall 2016  
Sonny Chan

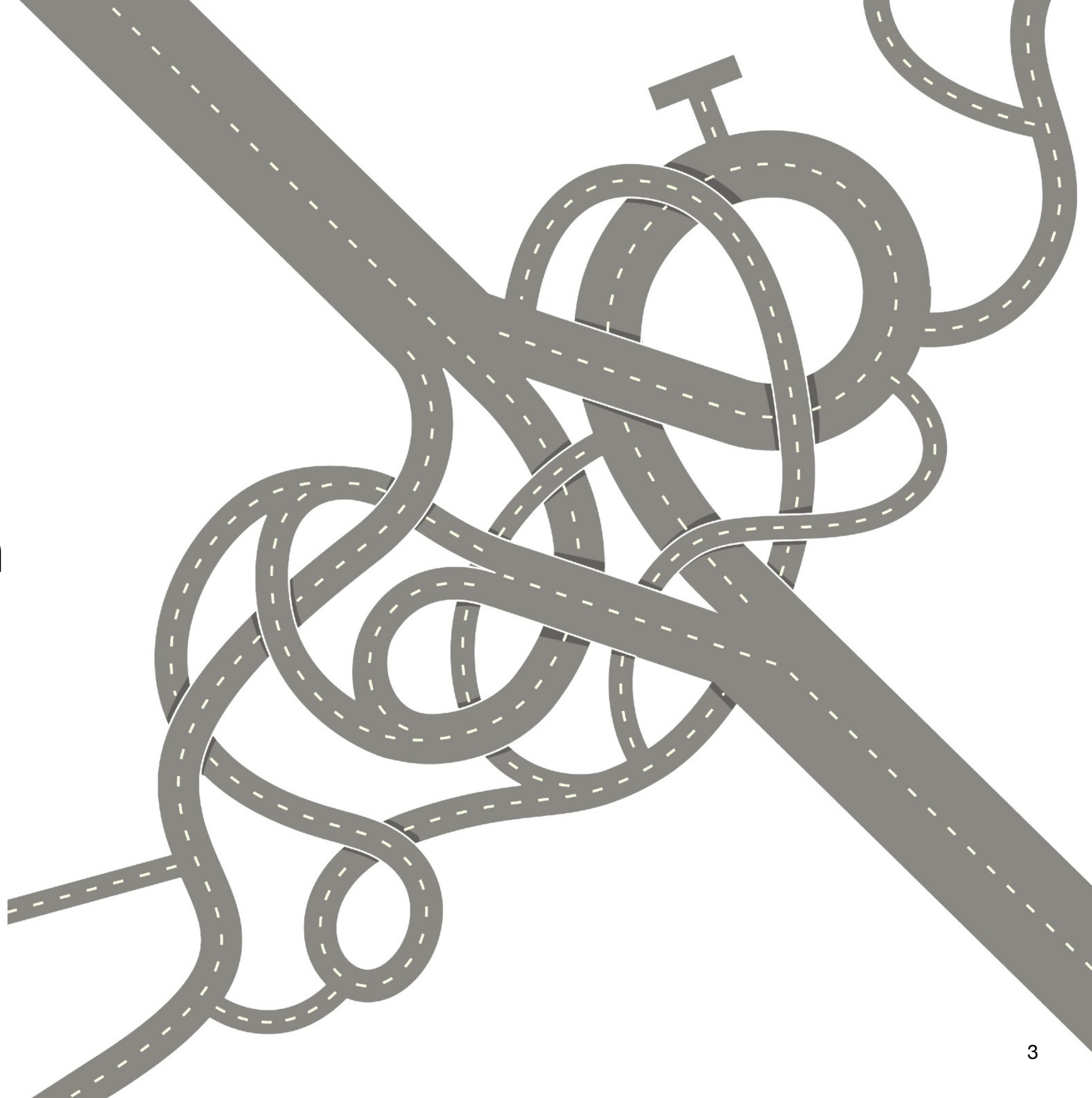


# Today's Outline

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- Convolution
  - discrete and continuous
  - one- and two-dimensional
- Convolution Filters on Images
  - edge detection
  - smoothing
- Colour Conversion

Convolution



Convolution is an operation on two functions.

$$f(x) \star g(x) \Rightarrow (f \star g)(x)$$

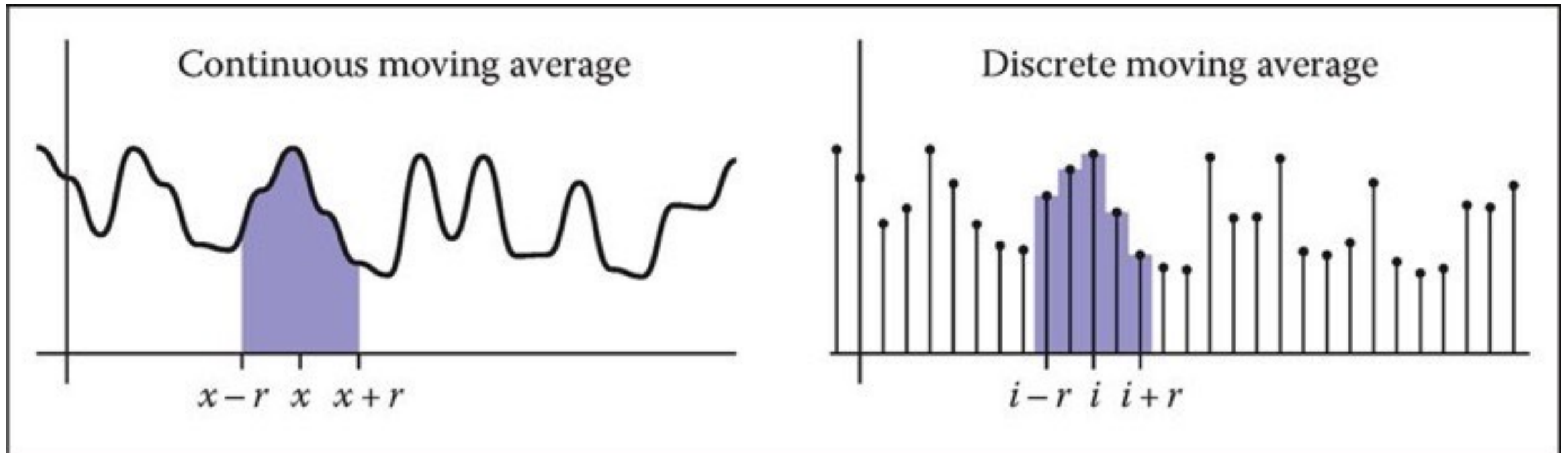
# Example: Moving Average

continuous function

$$f(x), x \in \mathbb{R}$$

discrete function

$$a[i], i \in \mathbb{Z}$$



$$h(x) = \frac{1}{2r} \int_{x-r}^{x+r} f(t) dt$$

$$c[i] = \frac{1}{2r+1} \sum_{j=i-r}^{i+r} a[j]$$

# Convolution Defined

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- Moving averages:

$$h(x) = \frac{1}{2r} \int_{x-r}^{x+r} f(t) dt$$

$$c[i] = \frac{1}{2r+1} \sum_{j=i-r}^{i+r} a[j]$$

- Convolution is a weighted moving average:

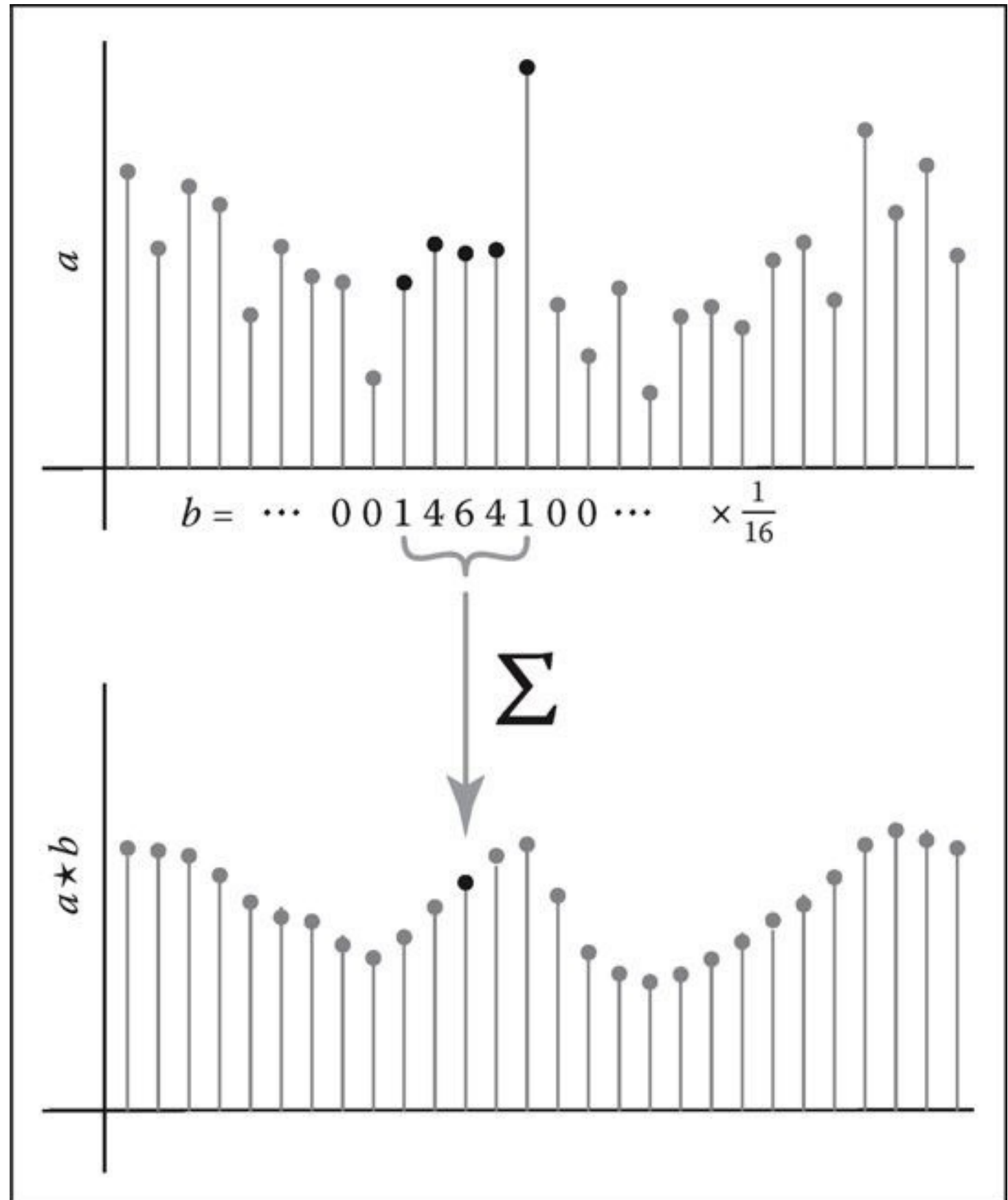
$$(f \star g)(x) = \int_{-\infty}^{\infty} f(t)g(x-t)dt$$

$$(a \star b)[i] = \sum_j a[j]b[i-j]$$

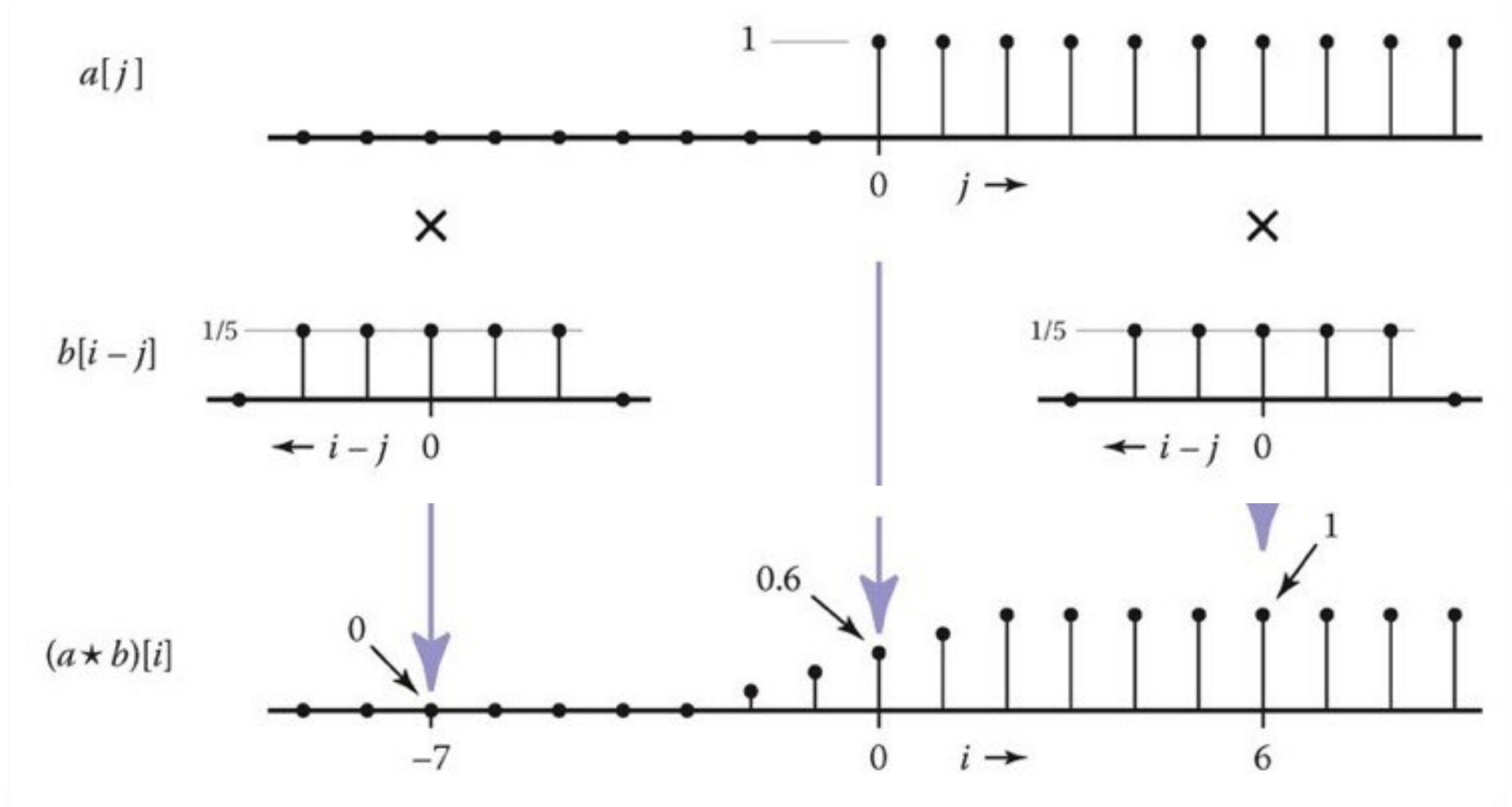
# Finite Support

- Most convolution *kernels* we use in the real world have *finite support*
- Second function is zero outside of a certain radius

$$(a \star b)[i] = \sum_{j=i-r}^{i+r} a[j]b[i-j]$$



# Example: Discrete Convolution with Box Filter





Is convolution  
**commutative?**

$$f \star g = g \star f$$

Is convolution  
**associative?**

$$(f \star g) \star h = f \star (g \star h)$$

Is convolution  
**distributive?**

$$(f + g) \star h = (f \star h) + (g \star h)$$

What is convolution's  
**identity?**

$$(f \star \delta)(x) = f(x)$$

Does convolution have an  
**inverse?**

$$(f \star f^{-1})(x) = \delta(x)$$

# Summary of Convolution Properties

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- Commutative  $f \star g = g \star f$
- Associative  $(f \star g) \star h = f \star (g \star h)$
- Distributive (over addition)  $(f + g) \star h = (f \star h) + (g \star h)$
- Identity  $(f \star \delta)(x) = f(x)$

# Discrete-Continuous Convolution

---

- If we have one function of each type:

$$\begin{aligned}a[i] &: \mathbb{Z} \mapsto \mathbb{R} \\ f(x) &: \mathbb{R} \mapsto \mathbb{R}\end{aligned}$$

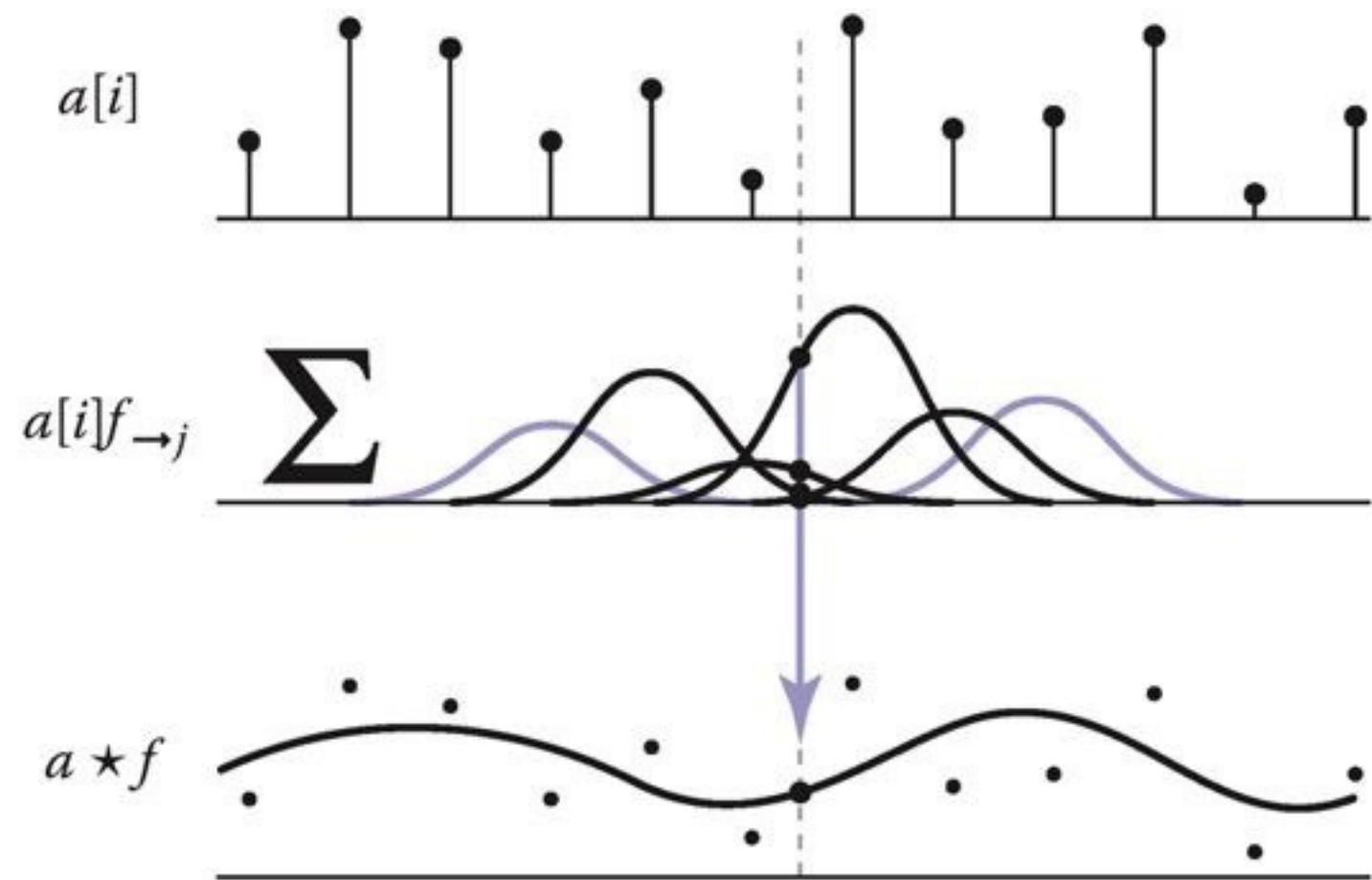
- We can define:

$$(a \star f)(x) = \sum_i a[i] f(x - i)$$

- When might we want to use this?

# Signal Reconstruction

by convolution with reconstruction  
kernel (e.g. cubic spline)





# Convolution in 2D

---

- If we have functions of two (or more) variables:

$$f(x, y), g(x, y) : \mathbb{R}^2 \mapsto \mathbb{R}$$

$$a[i, j], b[i, j] : \mathbb{Z}^2 \mapsto \mathbb{R}$$

- We can also define 2D convolution:

$$(a \star b)[i, j] = \sum_{i'} \sum_{j'} a[i', j'] b[i - i', j - j']$$

- Where might we apply this?

# Images!

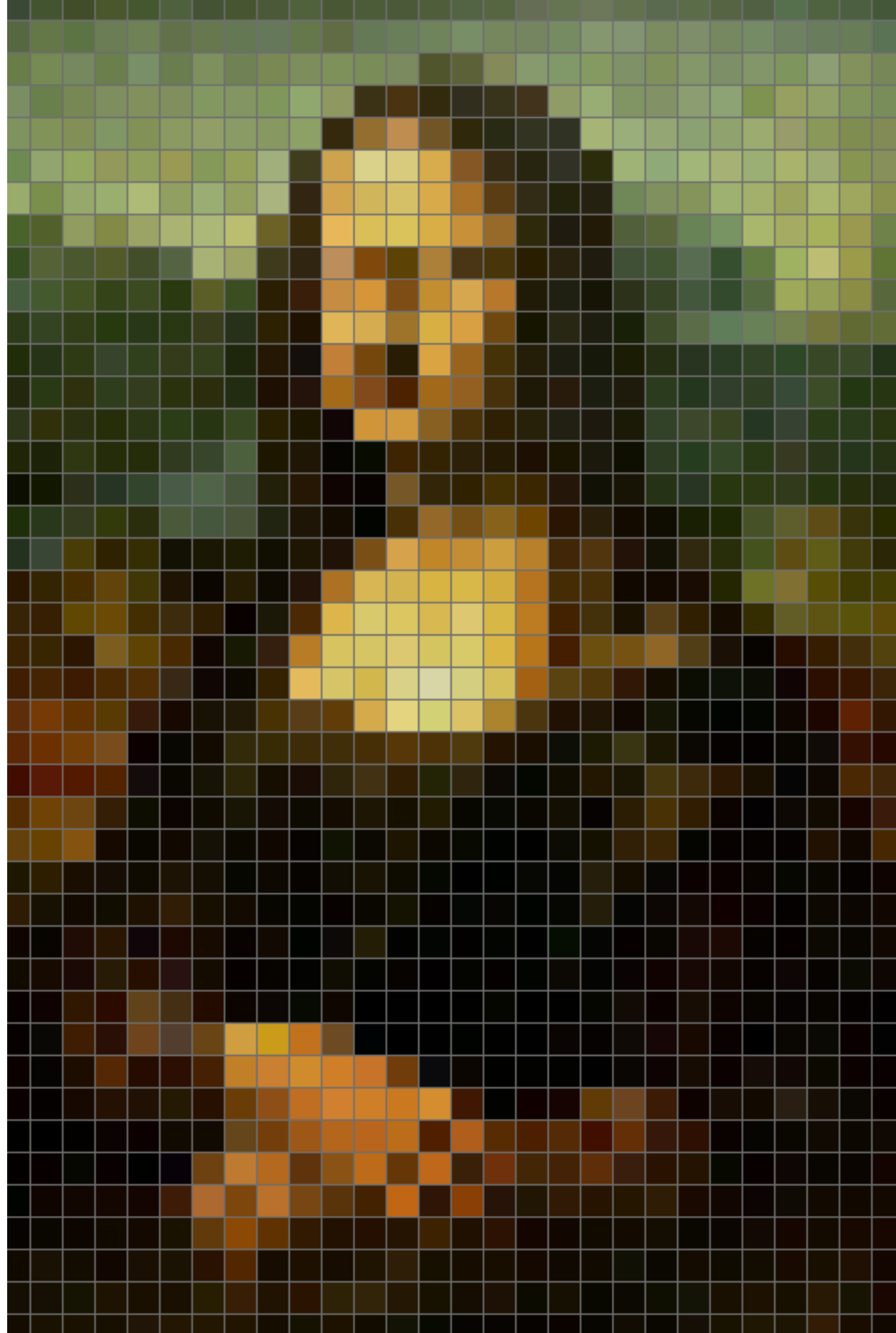
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- A sampled digital image is a discrete function of two variables:

$$I[i, j] : \mathbb{Z}^2 \mapsto \mathbb{R}$$

- Though, technically, for colour images it's like this:

$$I[i, j] : \mathbb{Z}^2 \mapsto \mathbb{R}^3$$

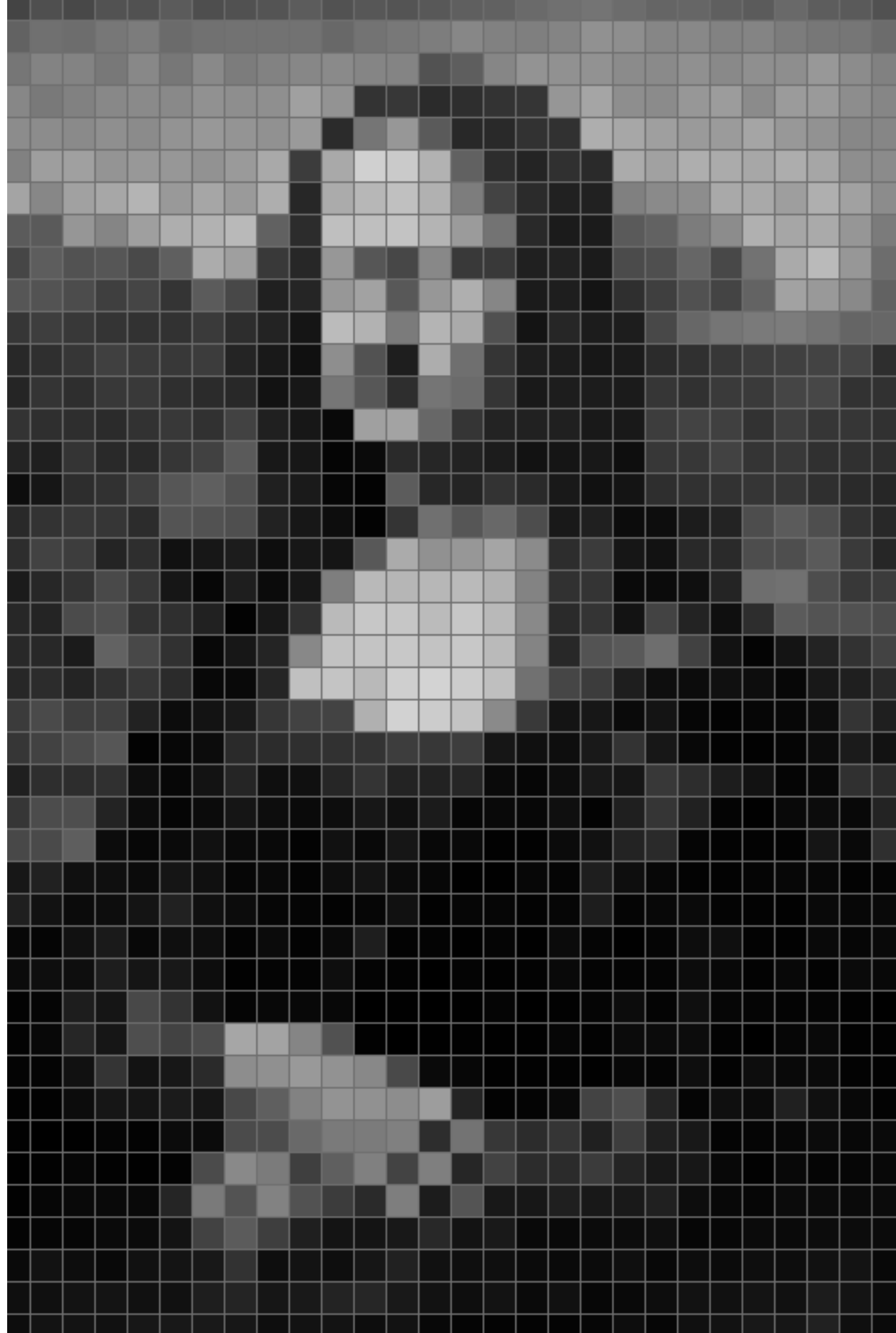


# Images!

---

- A sampled digital image is a discrete function of two variables:

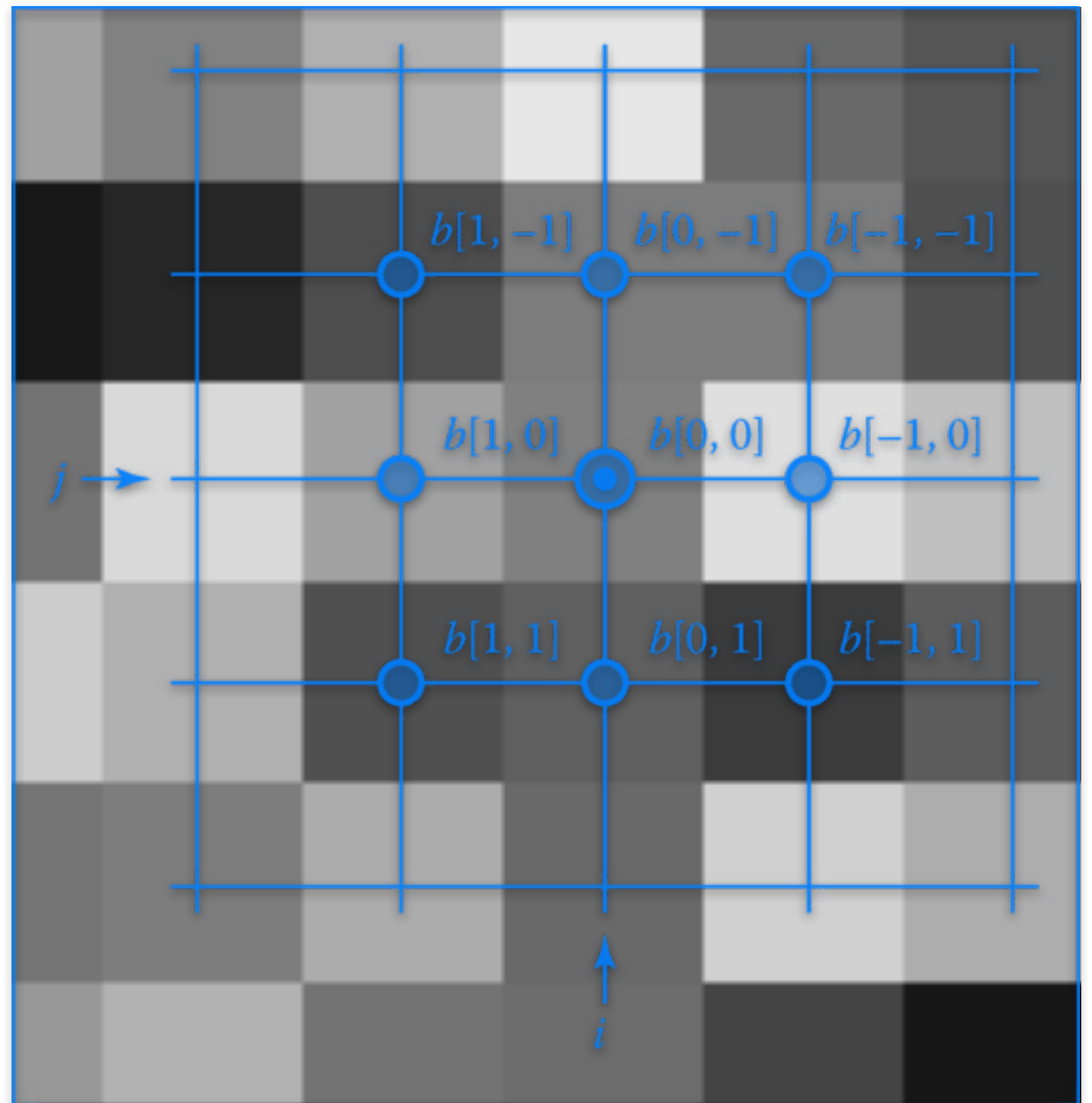
$$I[i, j] : \mathbb{Z}^2 \mapsto \mathbb{R}$$



# Example: Convolution with 2D Box Filter

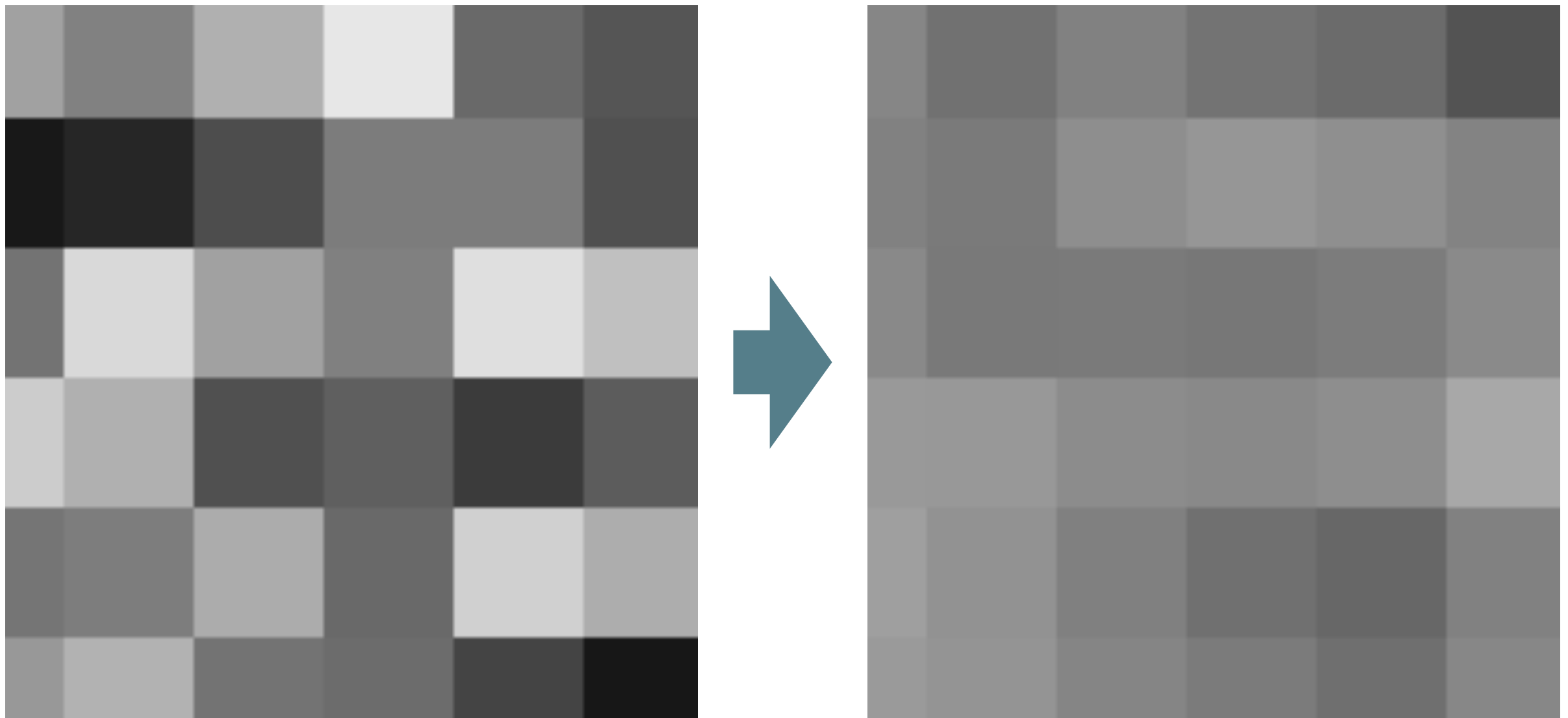
$$(a \star b)[i, j] = \sum_{i'} \sum_{j'} a[i', j'] b[i - i', j - j']$$

$$b[i, j] = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{9} & \frac{1}{9} & \frac{1}{9} & 0 \\ 0 & \frac{1}{9} & \frac{1}{9} & \frac{1}{9} & 0 \\ 0 & \frac{1}{9} & \frac{1}{9} & \frac{1}{9} & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$



# Convolution with 2D Box Filter

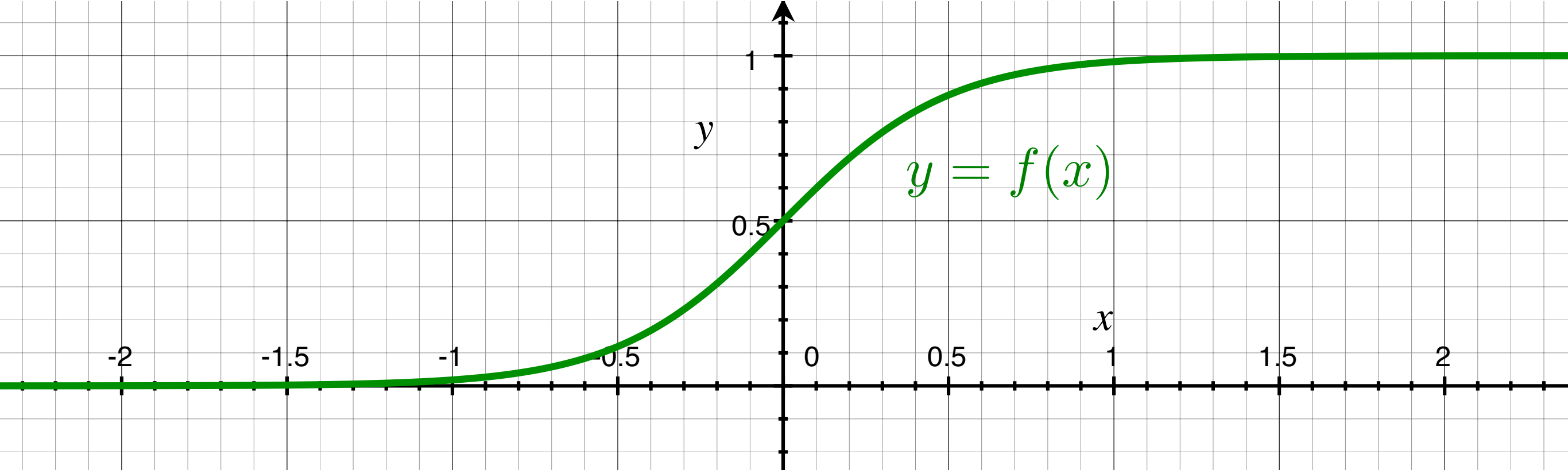
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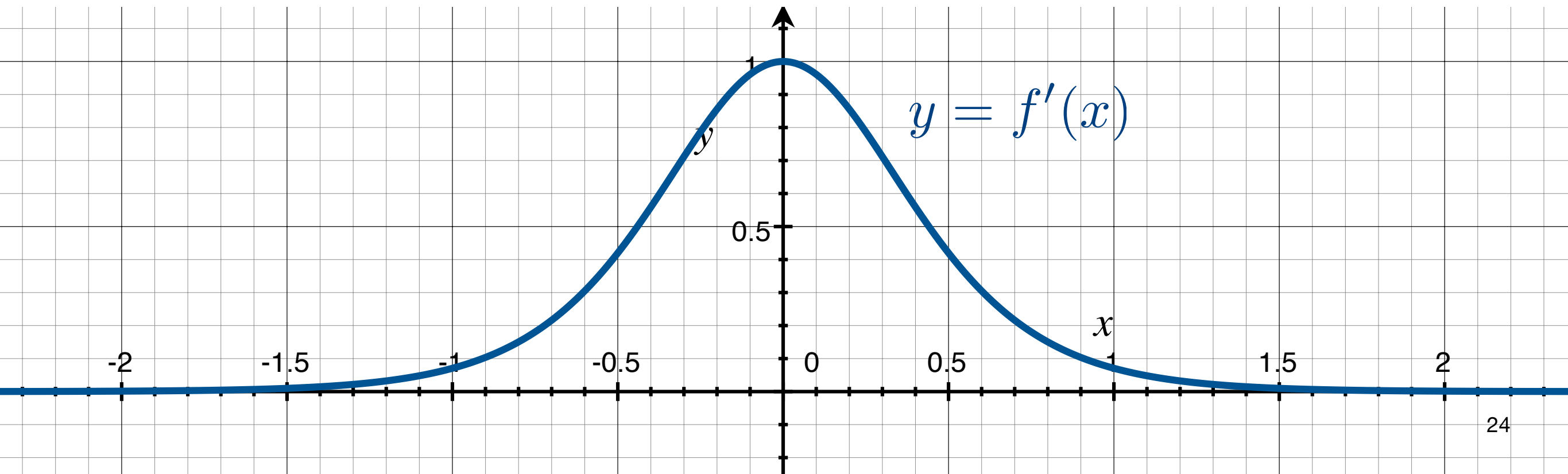
Can we achieve any  
**other effects?**

What constitutes  
**an edge?**





**How might we “detect” this edge?**





# An Edge Detector

---

- Derivative

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

- Forward difference

$$f'(x) \approx \frac{f(x+h) - f(x)}{h}$$

- Central difference

$$f'(x) \approx \frac{f(x + \frac{h}{2}) - f(x - \frac{h}{2})}{h}$$

- Discrete convolution

$$b[i] = [\dots 0 \quad +1 \quad 0 \quad -1 \quad 0 \dots]$$

# Edge Detection in 2D: Sobel Filters

---

$$G_x[i, j] = \begin{bmatrix} +1 & 0 & -1 \\ +2 & 0 & -2 \\ +1 & 0 & -1 \end{bmatrix}$$

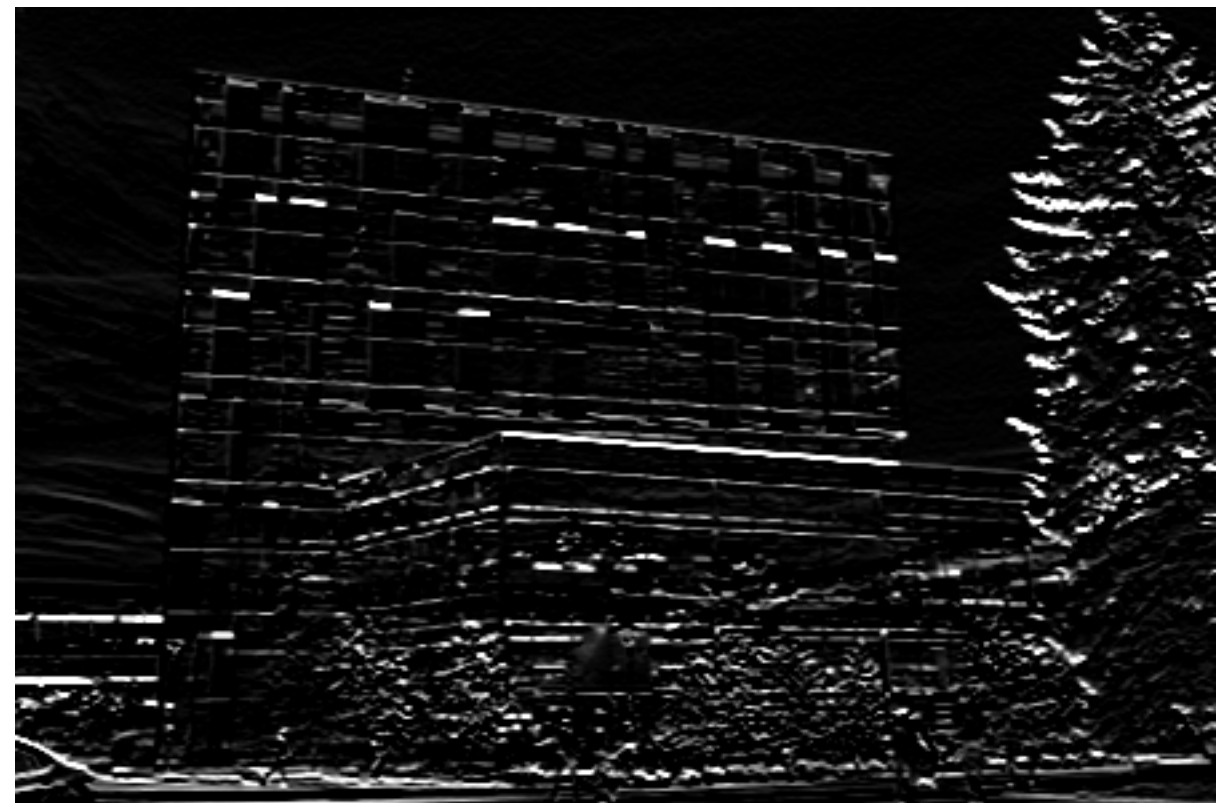
vertical edges

$$G_y[i, j] = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ +1 & +2 & +1 \end{bmatrix}$$

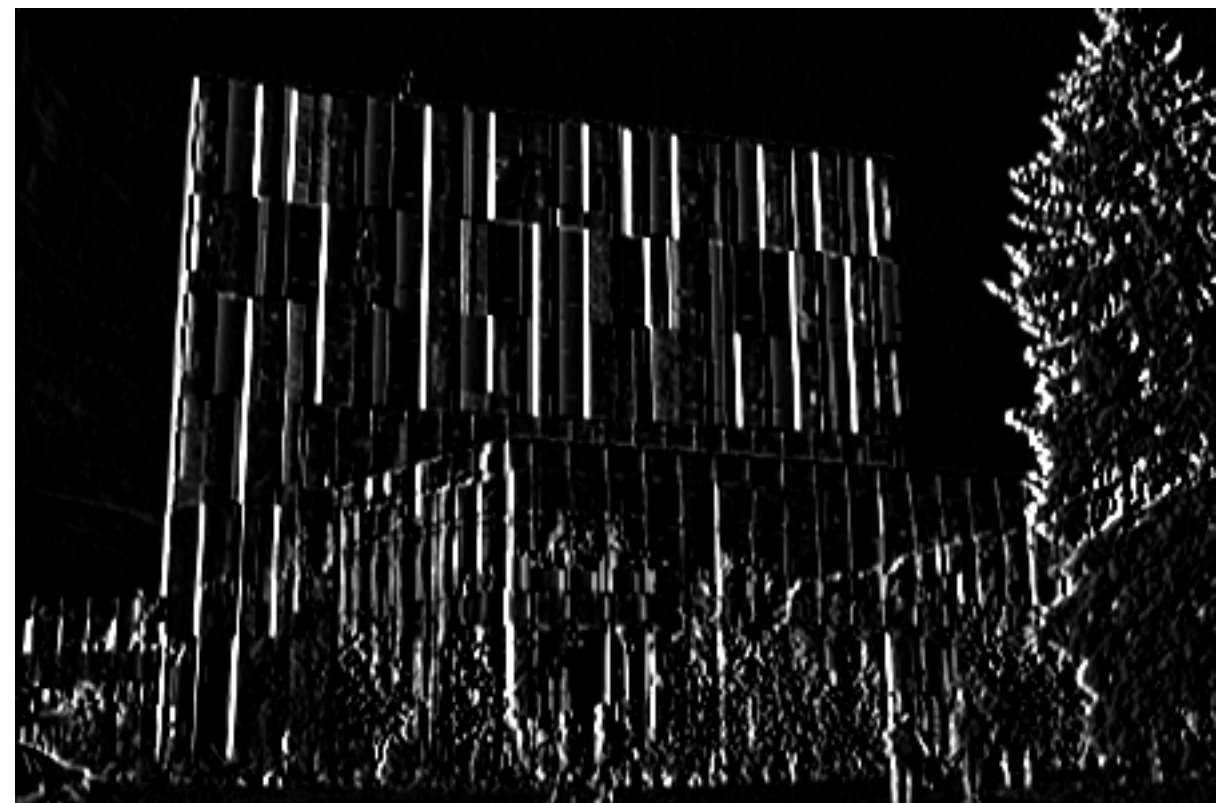
horizontal edges



original



horizontal edges



vertical edges

# Things to Remember

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- Convolution is an operation on two functions
  - can be applied on continuous or discrete functions
  - can be done in 1D, 2D, or more dimensions
  - continuous ★ discrete = continuous output function
- Convolution is used for many important things:
  - signal reconstruction (interpolation)
  - dazzling image effects, and more...