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# Image processing 101

Fourier Fun

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# Images in computers

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- Discretized in space  $\implies$  Pixels *picture elements*
- Discretized in intensity/color
  - Resolution: 8-bit, 14-bit, 16-bit ...
  - Color representation: RGB (additive: red green blue), CMY (subtractive: cyan magenta yellow), HSV (hue saturation brightness; color wheel)
  - Color: quantized palette (8-bit) or true color (24-bit)
- Registration:
  - Photographic film + photometer
  - vacuum tube (TV camera)
  - solid-state devices (CCD, CID)
- Transmission: Serialization
  - Bandwidth limitations

# Linear Bases

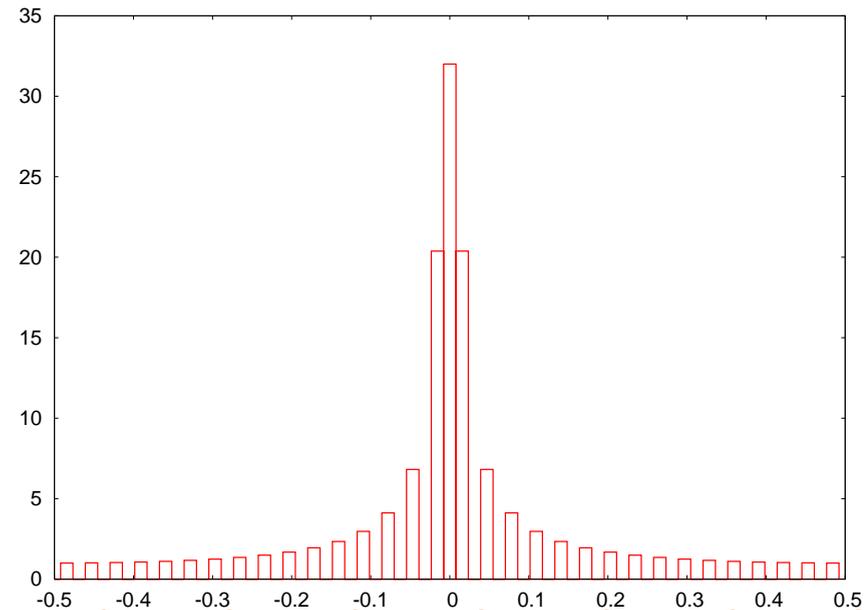
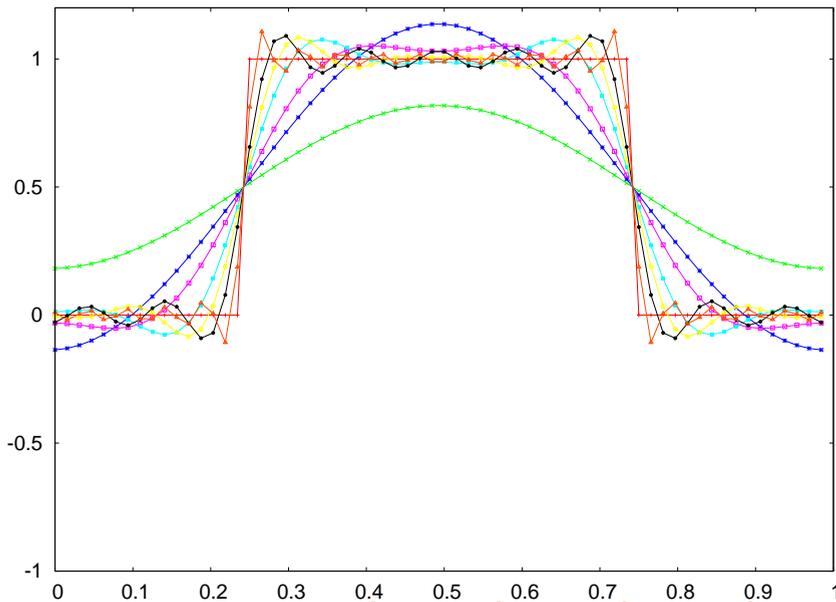
Linear decomposition of functions:

$$f(x) = \sum_k c_k b_k(x)$$

Fourier coefficient  
Fourier basis function

Fourier series expansion:

$$b_k(x) = e^{ik \cdot x}$$



# Plane waves and complex numbers

Complex numbers

$$C = \operatorname{Re} C + i \operatorname{Im} C = |C| e^{i\phi} = |C| (\sin \phi + i \cos \phi)$$

norm      phase

$$C^* = \operatorname{Re} C - i \operatorname{Im} C$$

$$|C| = \sqrt{C^* C} = \sqrt{(\operatorname{Re} C)^2 + (\operatorname{Im} C)^2}$$

Plane waves

$$b_k(x) = e^{ik \cdot x} = \sin k \cdot x + i \cos k \cdot x$$

$$|b_k(x)| \equiv 1$$

Interference

$$b_{k_1}(x) + b_{k_2}(x) = e^{ik_1 \cdot x} + e^{ik_2 \cdot x}$$

$$|b_{k_1}(x) + b_{k_2}(x)|^2 = 2 + 2 \cos[(k_1 - k_2) \cdot x]$$

# The Fourier transform

Fourier synthesis:

$$f(x) = \frac{1}{N} \sum_k \hat{f}(k) e^{ik \cdot x}$$

Fourier analysis:

$$\begin{aligned} \hat{f}(k) = (\mathcal{F}f)(k) &= \sum_x f(x) e^{-ik \cdot x} \\ &= \langle f, b_k^* \rangle \end{aligned}$$

↙ Scalar product

$k$  is the *wavenumber*: Frequency  $\nu$

$$\nu = \frac{k}{2\pi}$$

Linearity:

$$\begin{aligned} \mathcal{F}(f + g) &= \mathcal{F}f + \mathcal{F}g \\ \mathcal{F}(\alpha f) &= \alpha \mathcal{F}f \end{aligned}$$

# The Fourier transform

- Two different, complete representation of the same signal
  - Position/time space  
Localization of signals
  - Momentum/frequency space  
Repetition/Periodicity and shape  
Translation-invariant basis

- Power spectrum

$$\sum_x |f(x)|^2 = \sum_k |\hat{f}(k)|^2$$

- Fast-fourier transform
  - $O(N \log N)$

# Fourier modes and reciprocity

Discretization:

$$x = 0, \Delta x, 2\Delta x, \dots, L - 2\Delta x, L - \Delta x$$

Periodicity:

$$f(x) = f(x + L)$$

For plane waves:

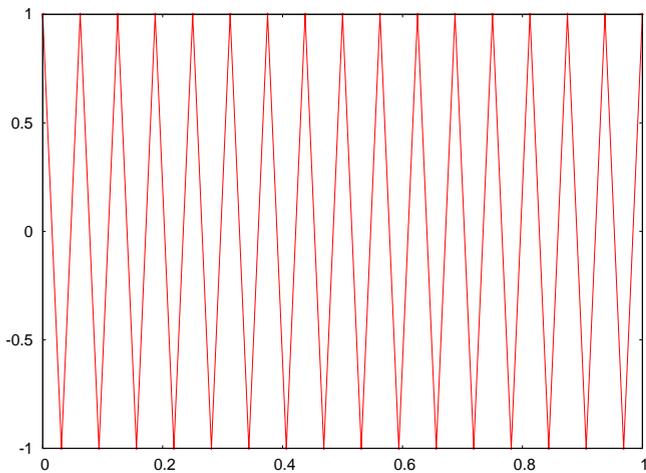
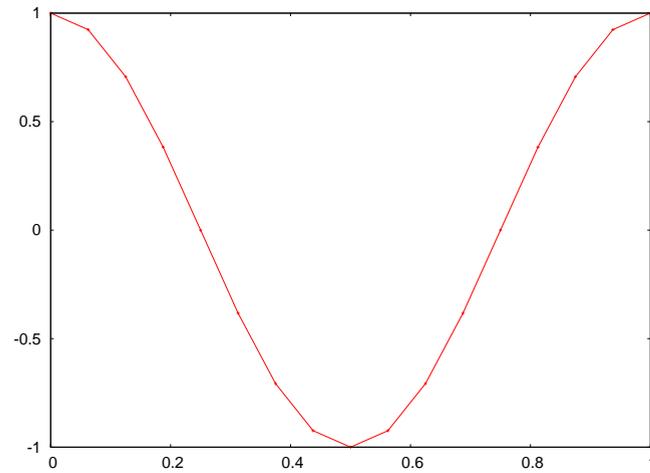
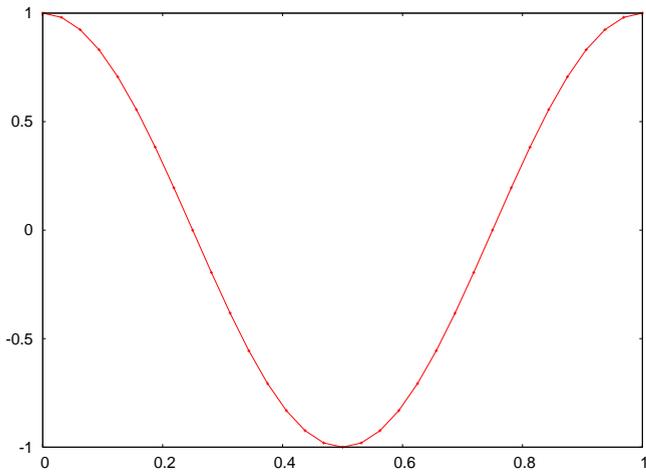
$$e^{ikx} = e^{ik(x+L)} = e^{ikx} e^{ikL} \quad \Rightarrow \quad e^{ikL} = 1$$

$$kL = 2\pi n \quad \Rightarrow \quad k = \frac{2\pi}{L} n$$

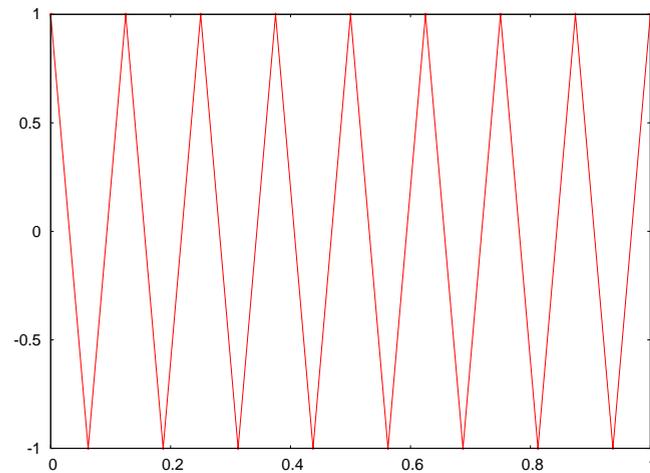
Highest-possible frequency:  $(-1, +1, -1, +1, \dots)$

$$(-1)^{x/\Delta x} = e^{-\pi(x/\Delta x)} \quad \Rightarrow \quad k_{\max} = \frac{\pi}{\Delta x} = \frac{2\pi}{L} \frac{N}{2} \quad \nu_{\max} = \frac{1}{2\Delta x}$$

# Fourier modes and reciprocity



$$L = 1, \Delta x = \frac{1}{32}$$

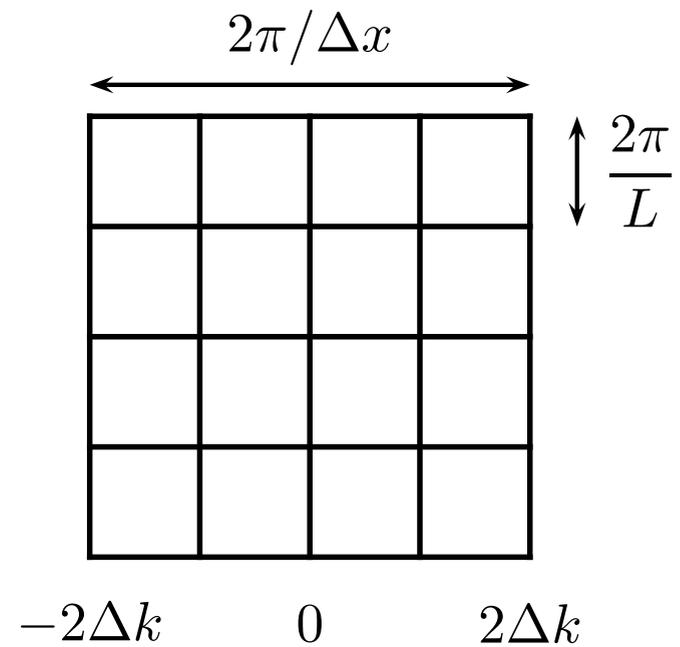
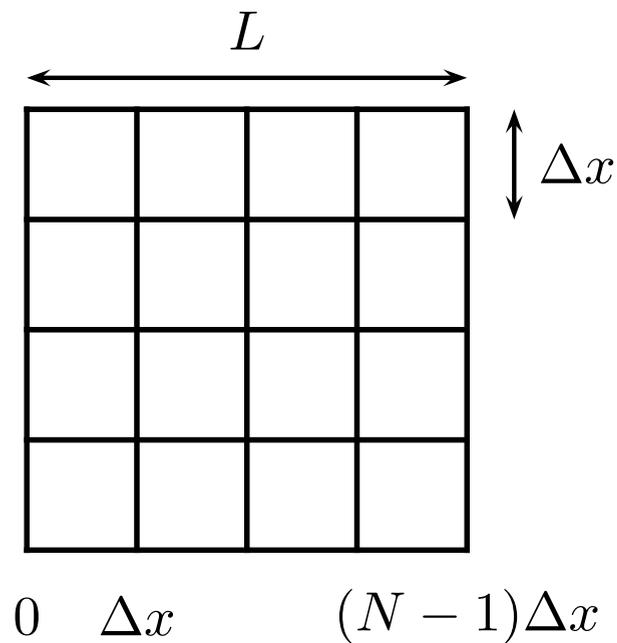


$$L = 1, \Delta x = \frac{1}{16}$$

# Dual lattices

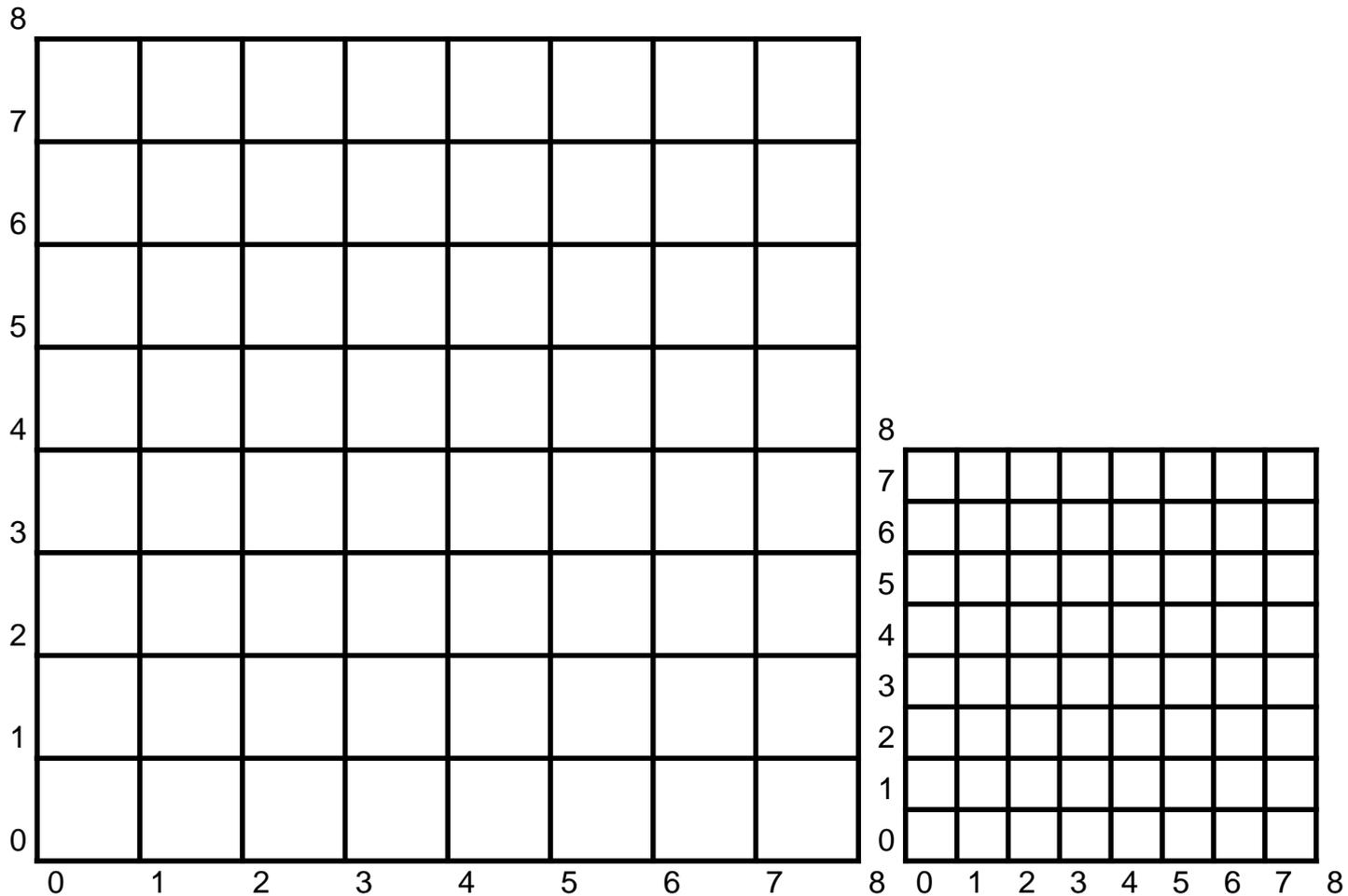
$$\Delta k = \frac{2\pi}{L}$$

$$x = 0, \Delta x, \dots, L - \Delta x \quad k = -\Delta k \frac{N}{2}, \dots, \Delta k \left( \frac{N}{2} - 1 \right)$$



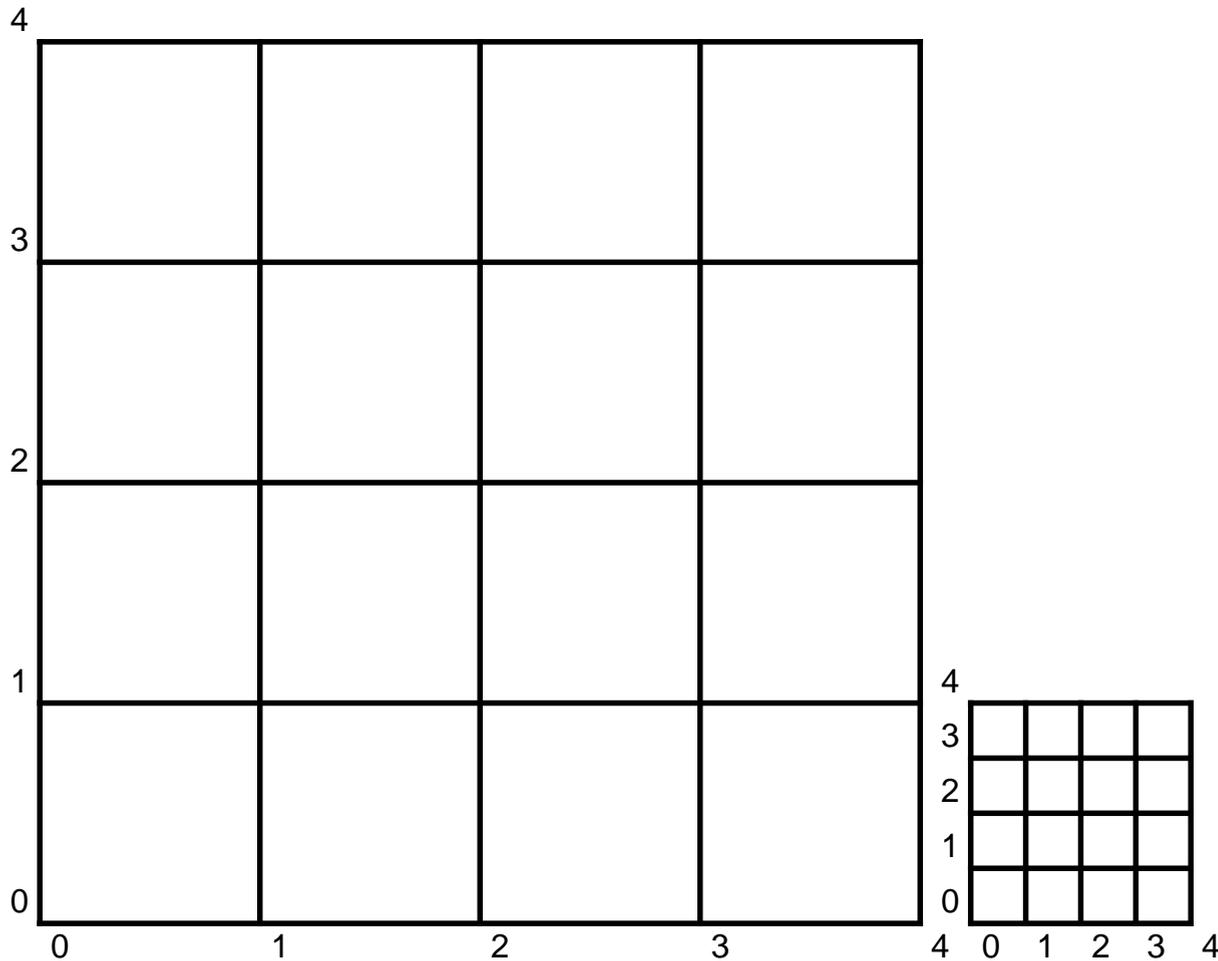
# Dual lattices

$$L \longrightarrow 2L \quad \Rightarrow \quad \Delta k \longrightarrow \frac{\Delta k}{2}$$

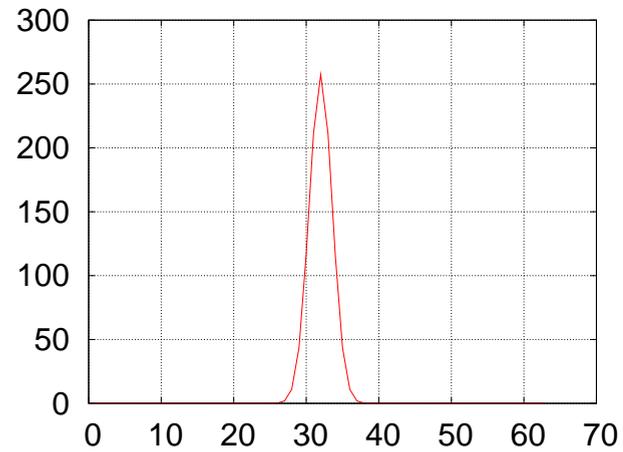
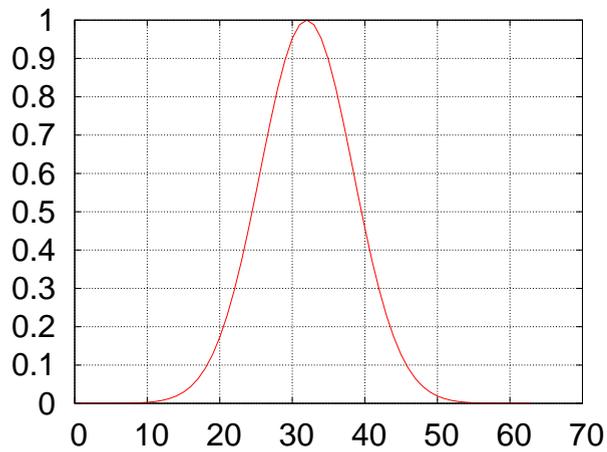
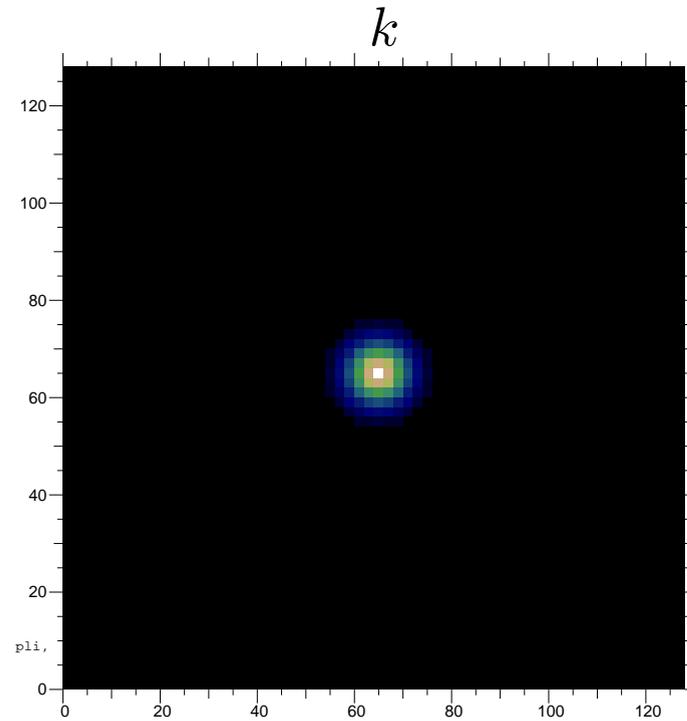
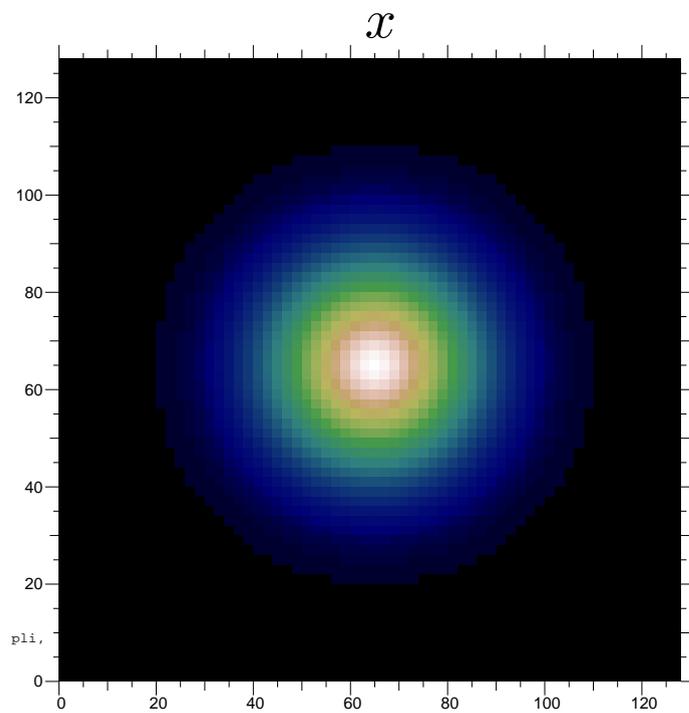


# Dual lattices

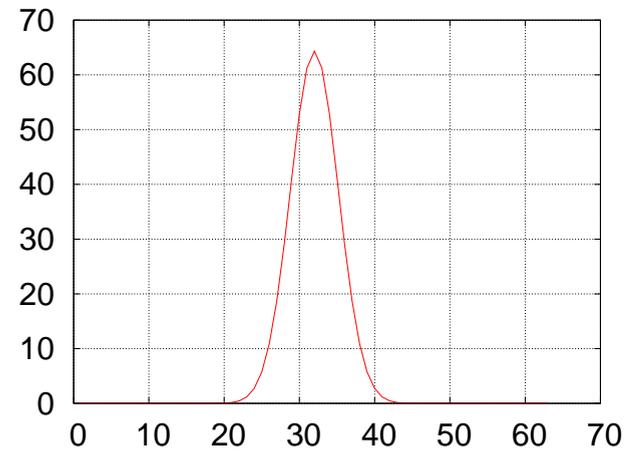
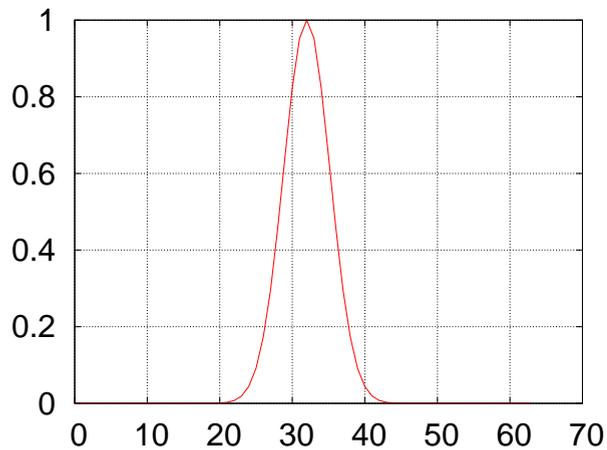
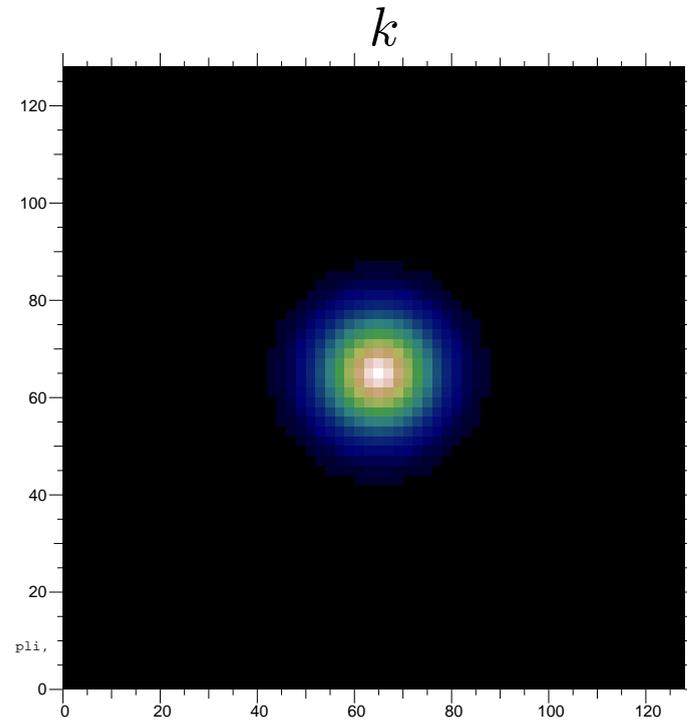
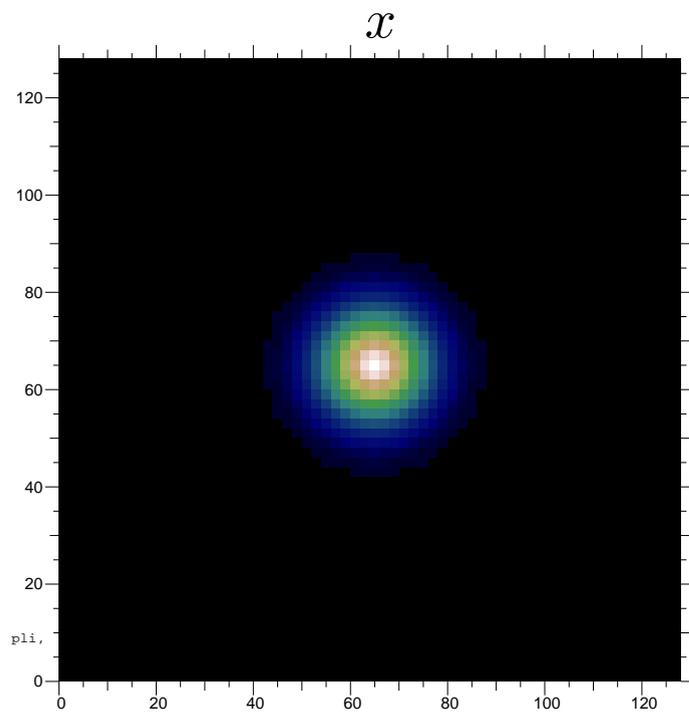
$$\Delta x \longrightarrow 2\Delta x \quad \Rightarrow \quad k_{\max} \longrightarrow \frac{k_{\max}}{2}$$



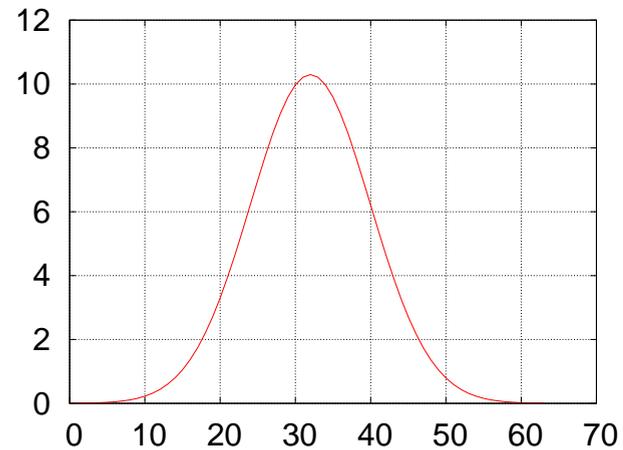
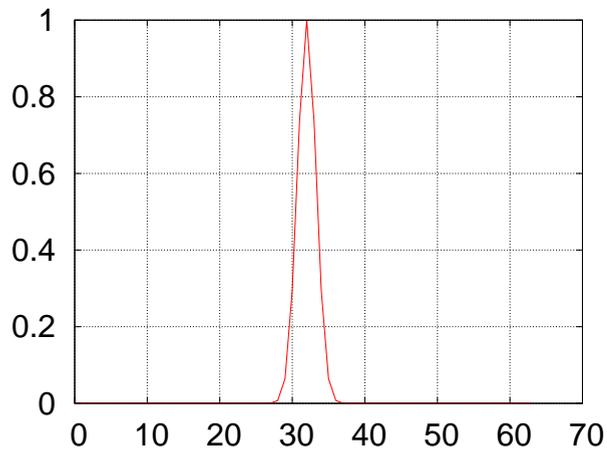
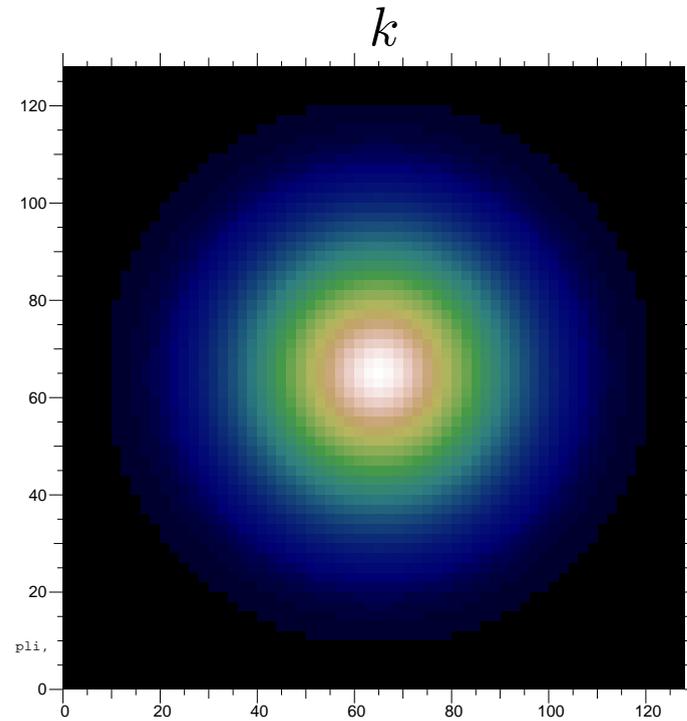
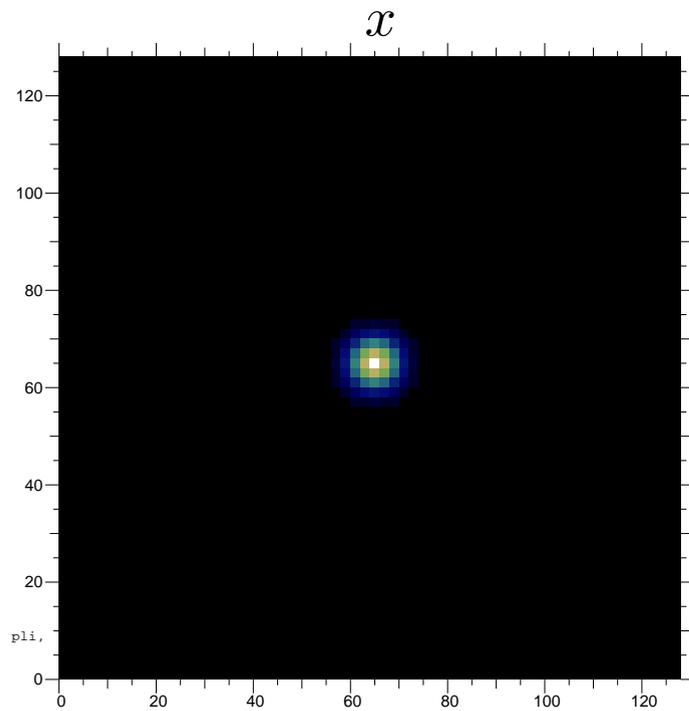
# Blob in Fourier space



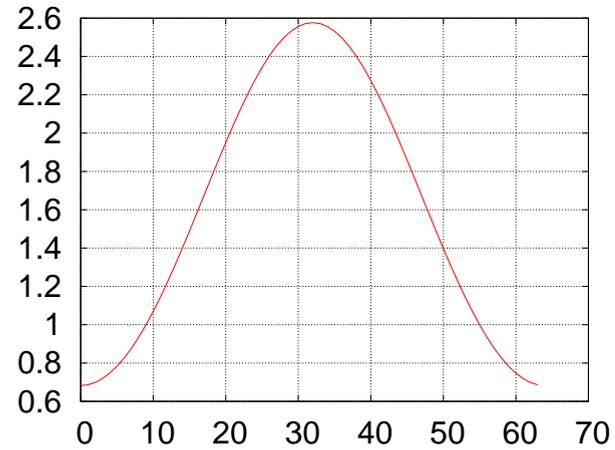
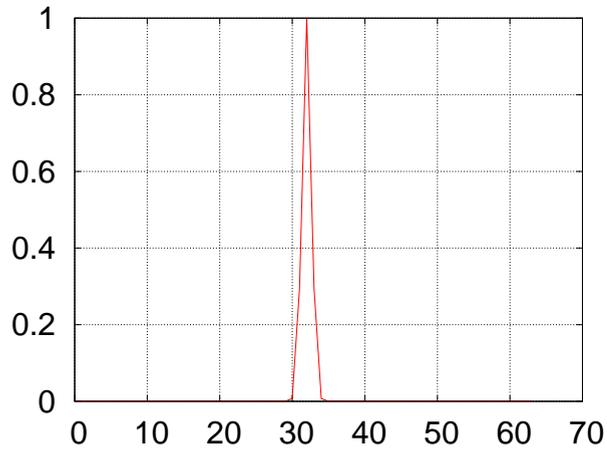
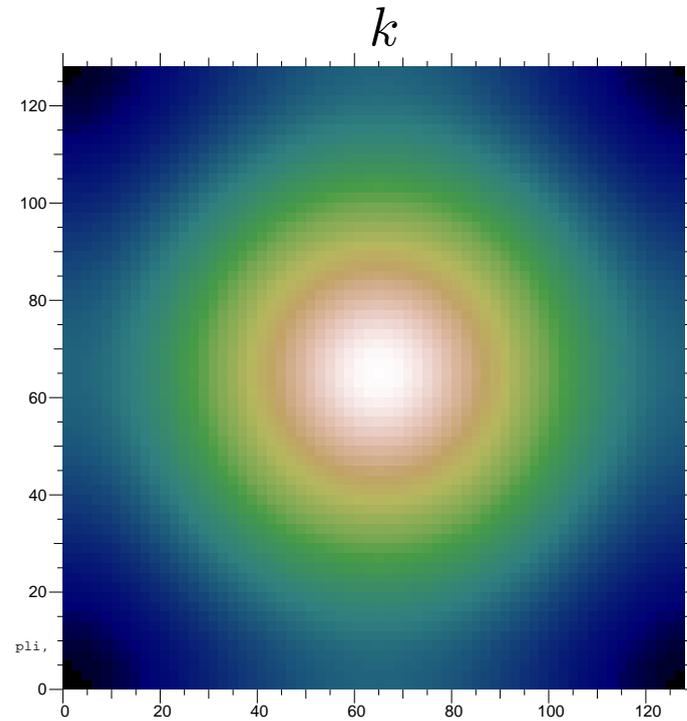
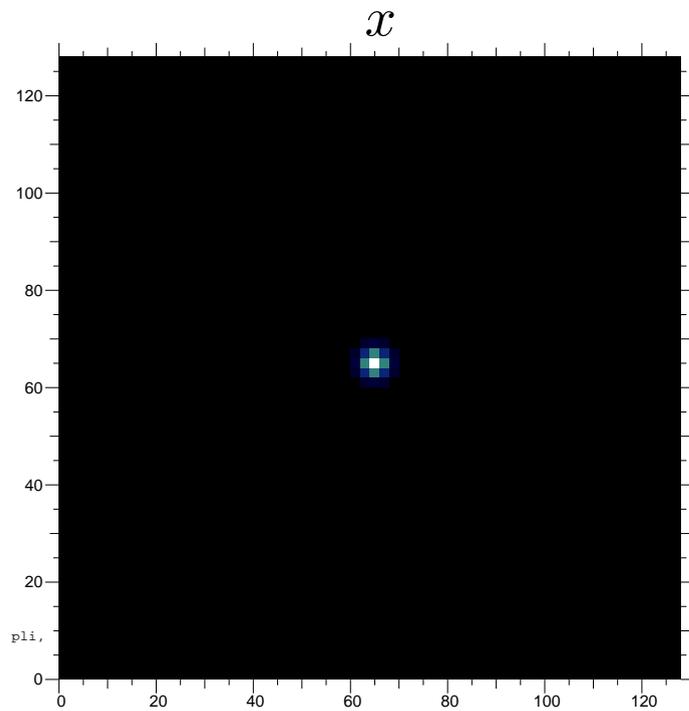
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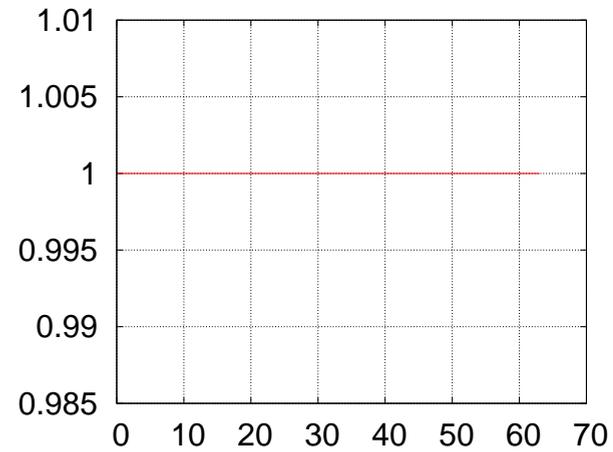
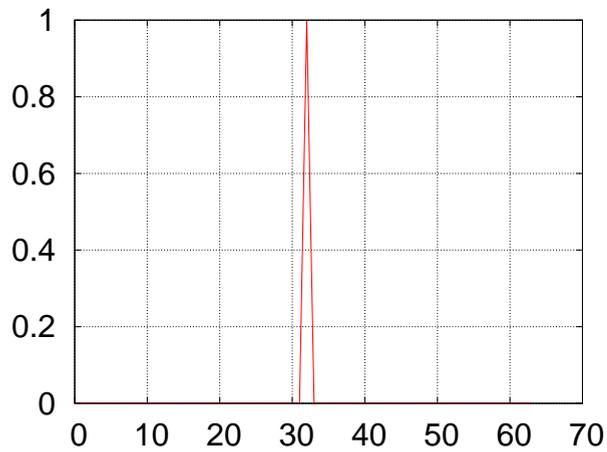
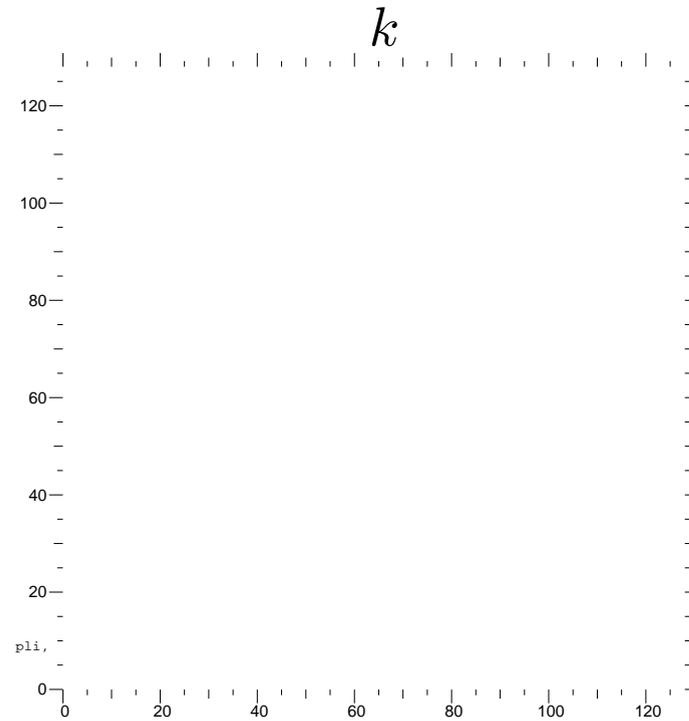
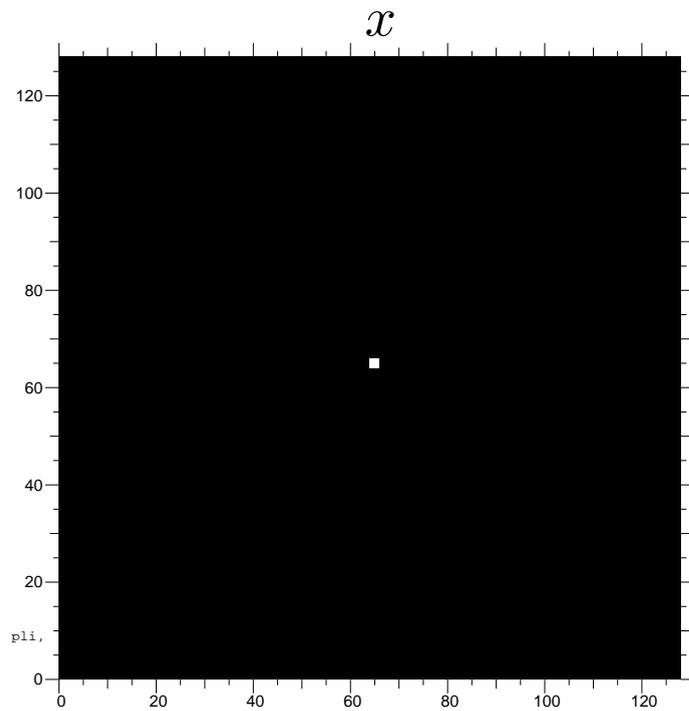
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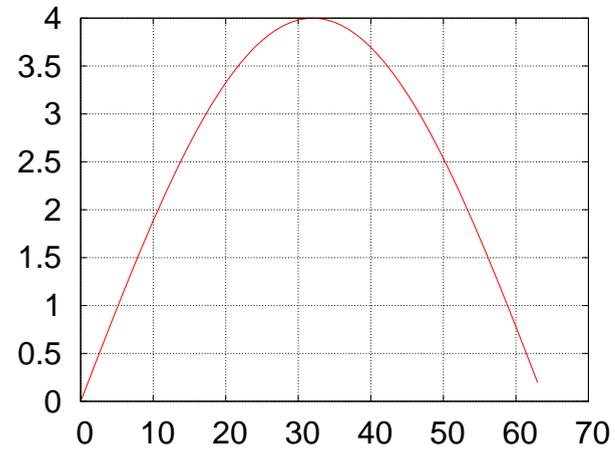
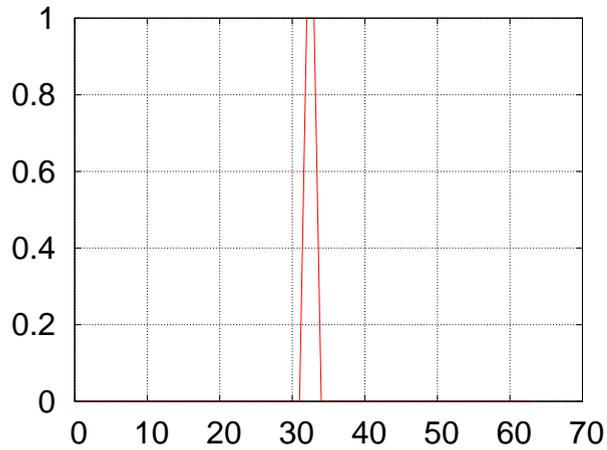
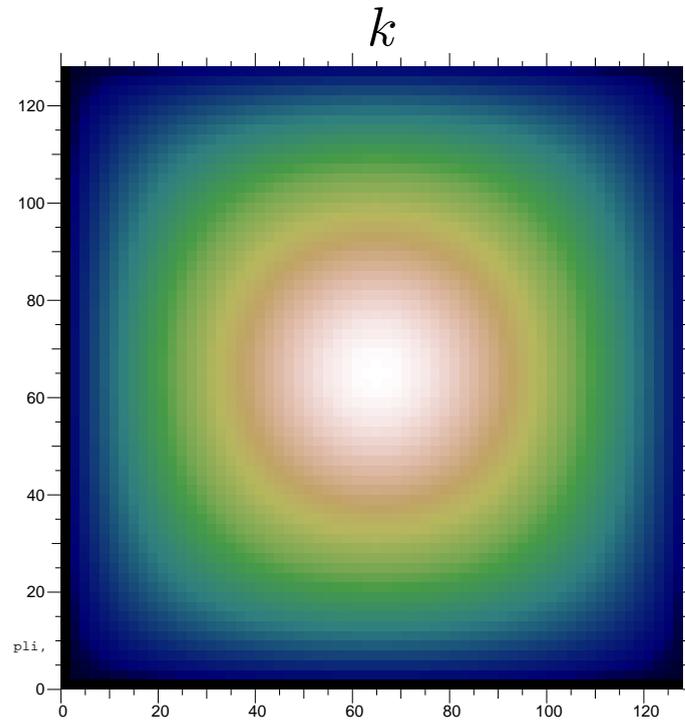
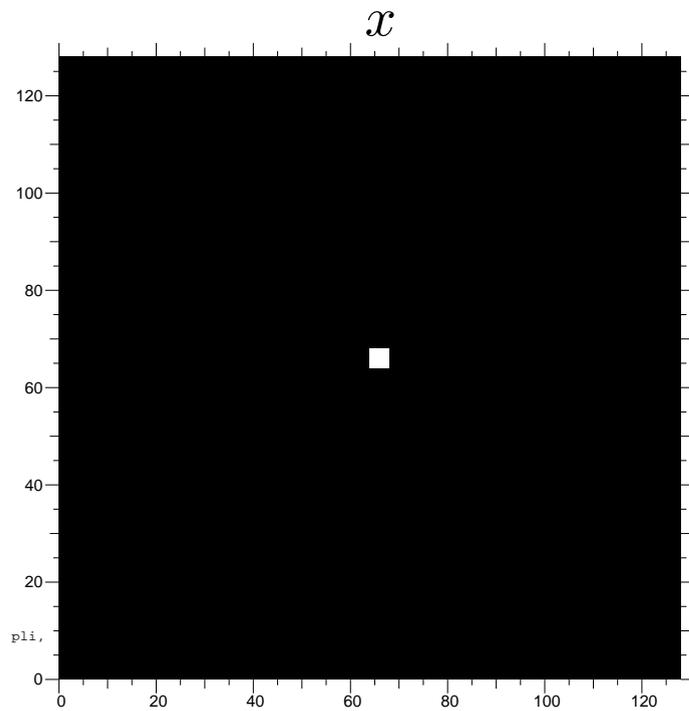
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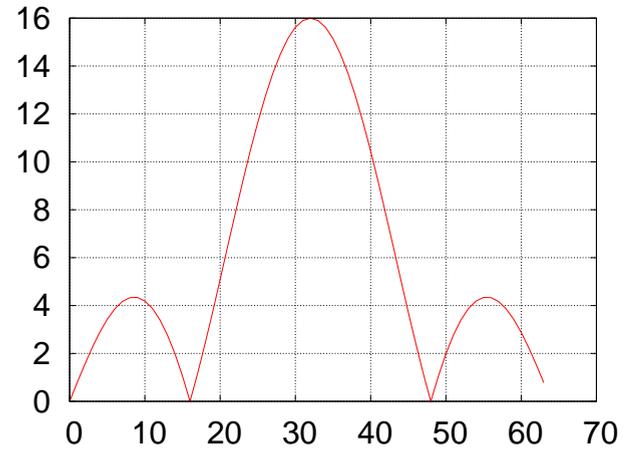
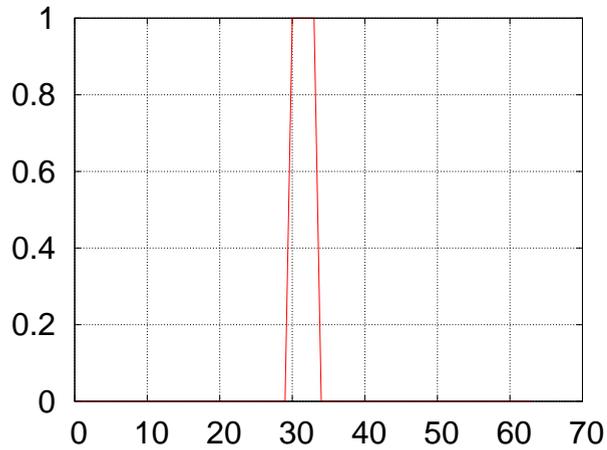
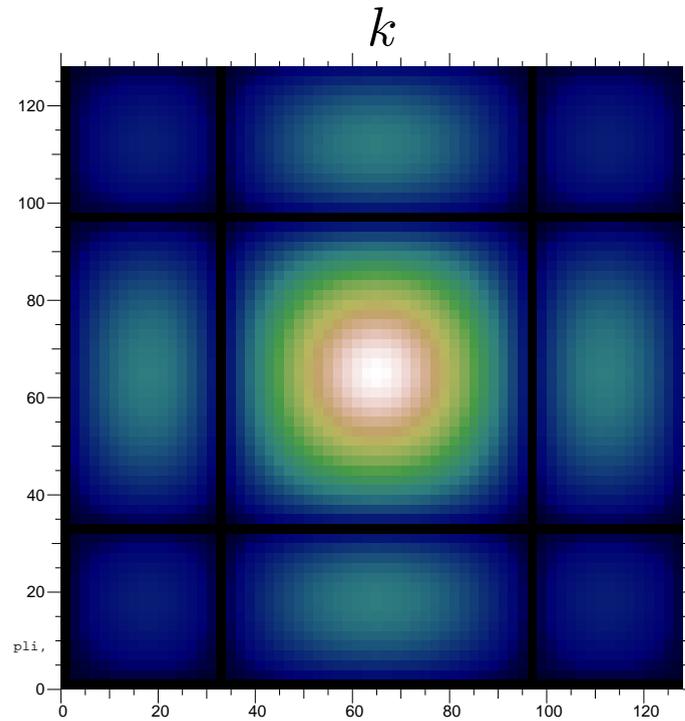
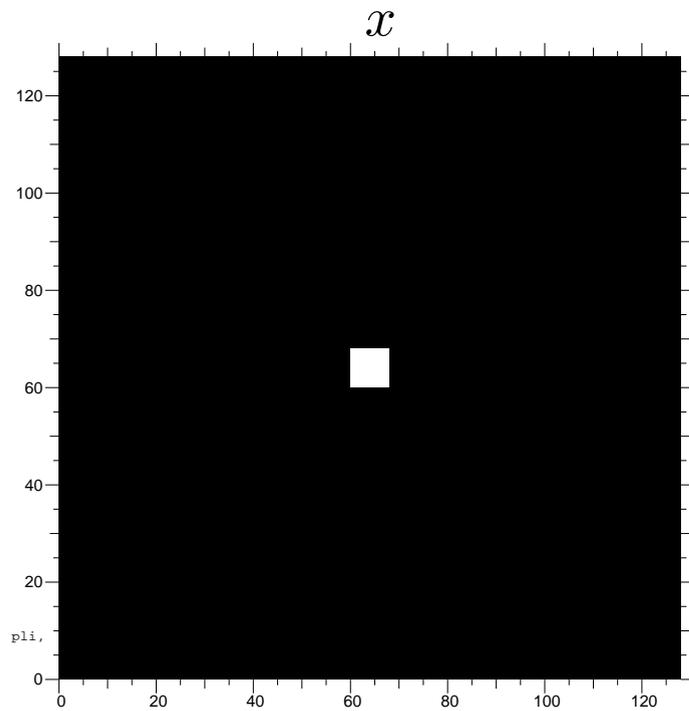
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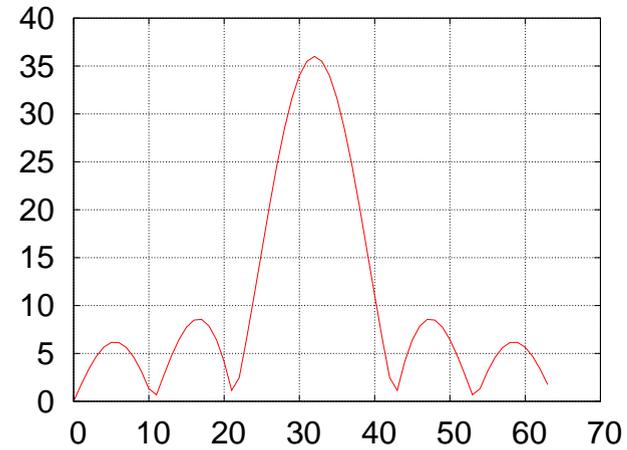
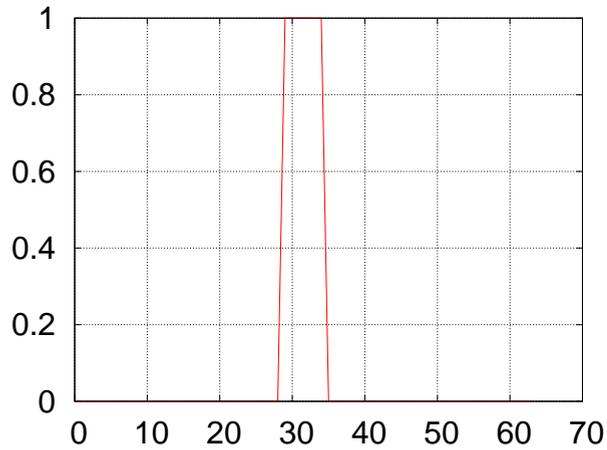
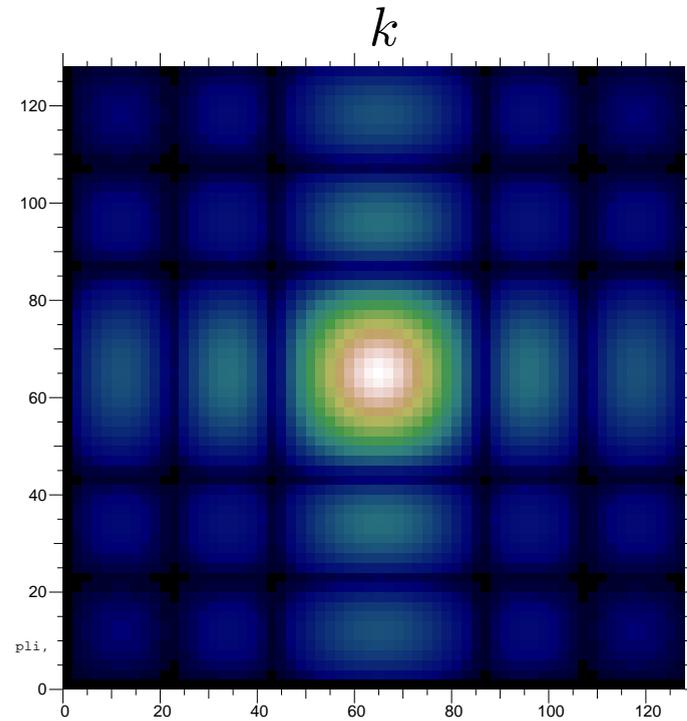
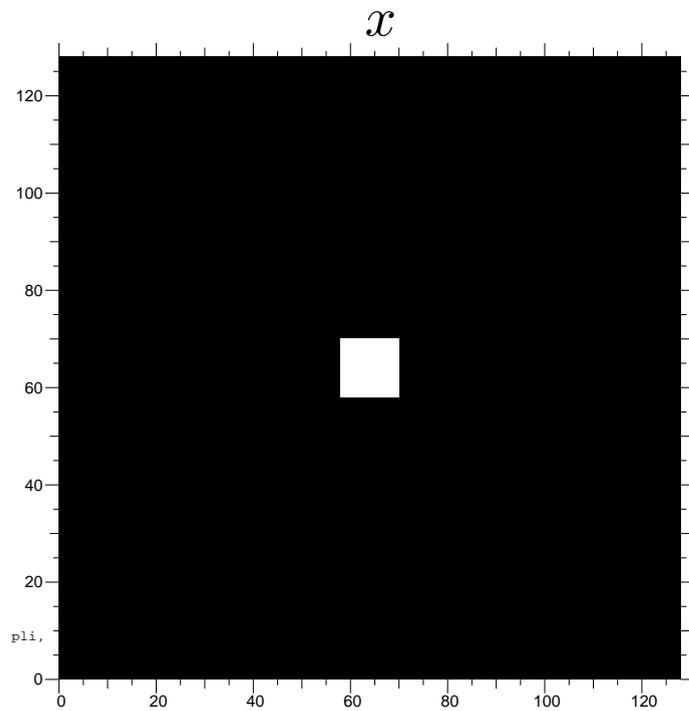
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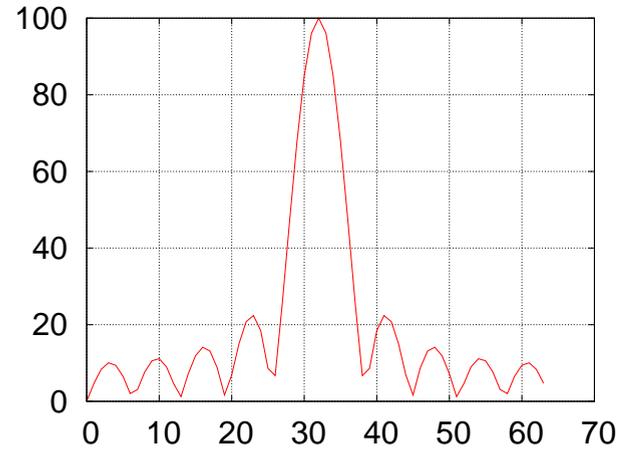
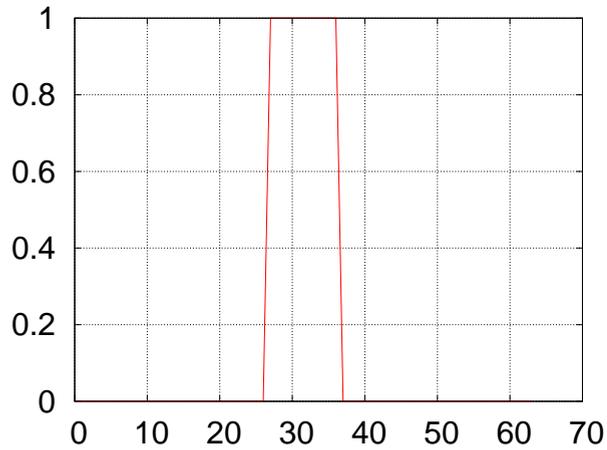
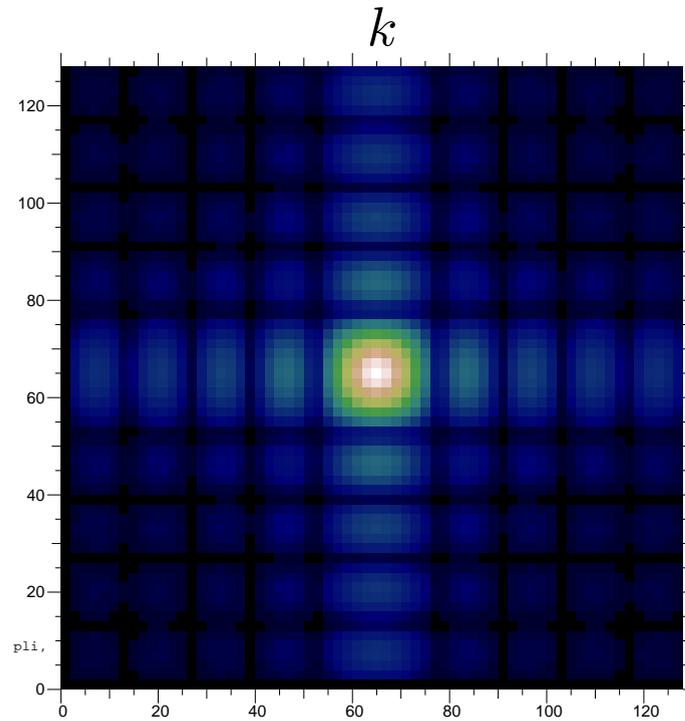
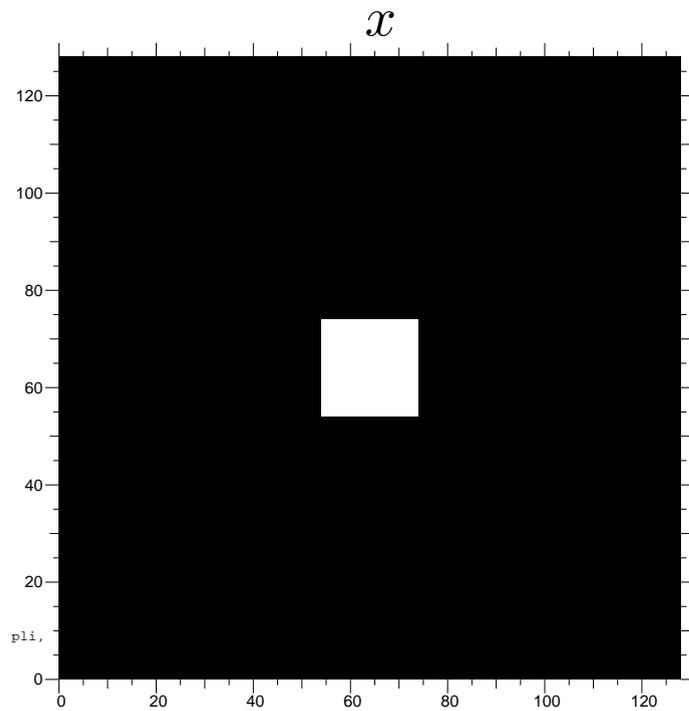
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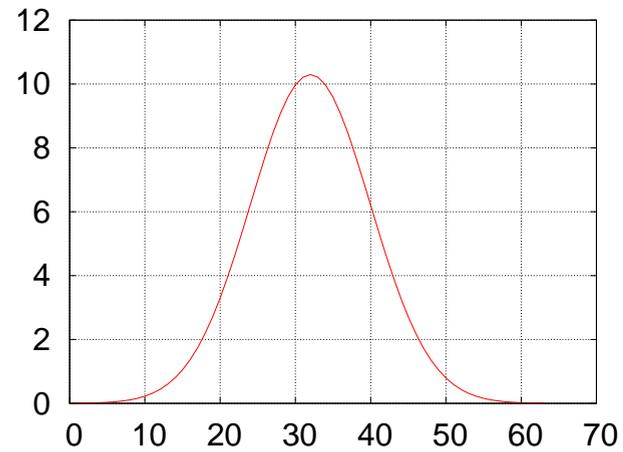
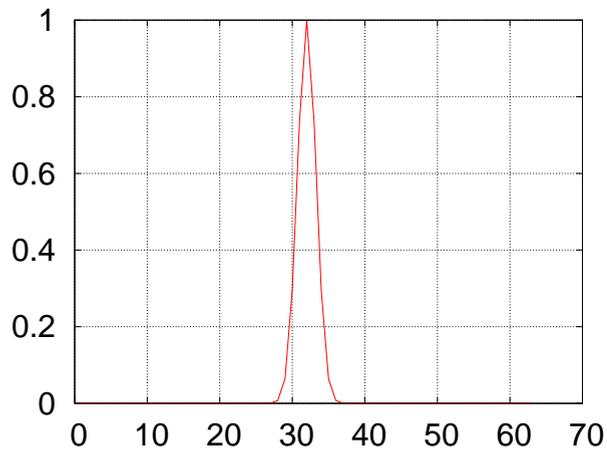
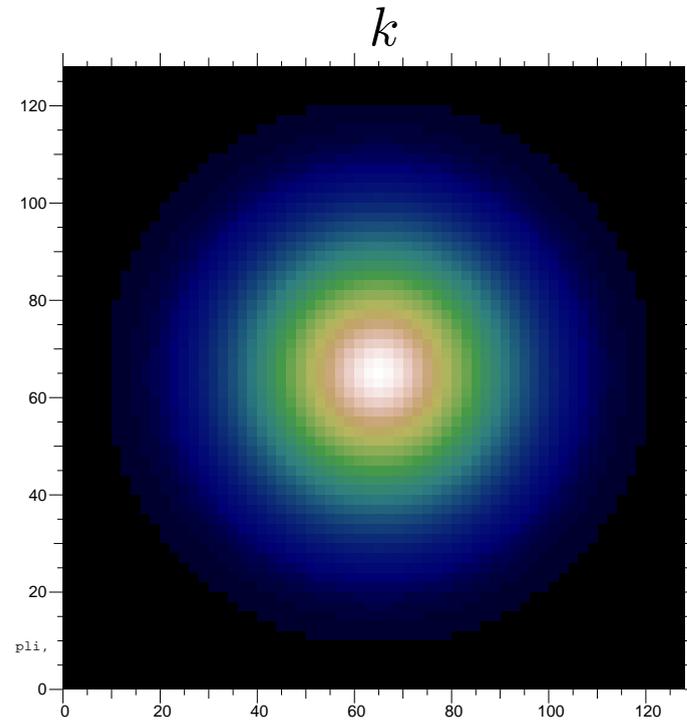
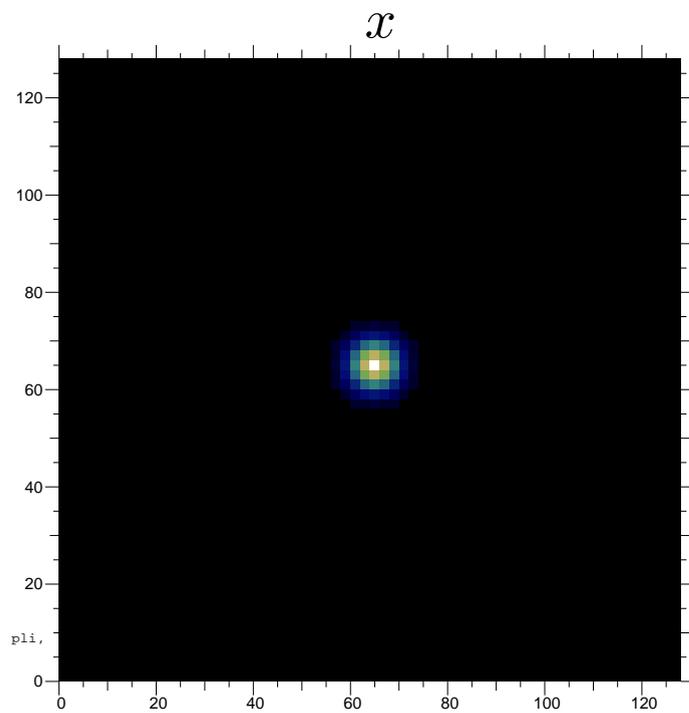
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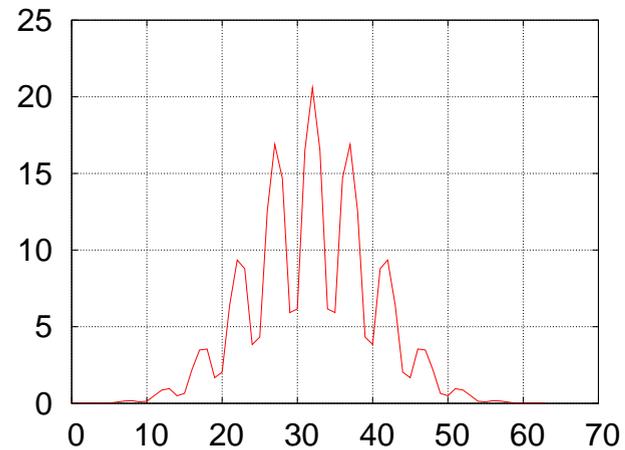
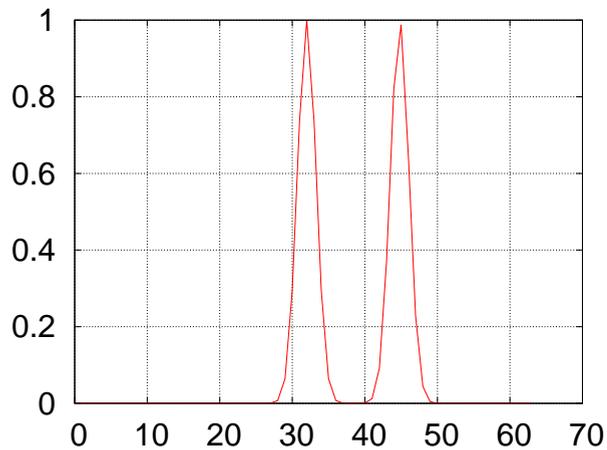
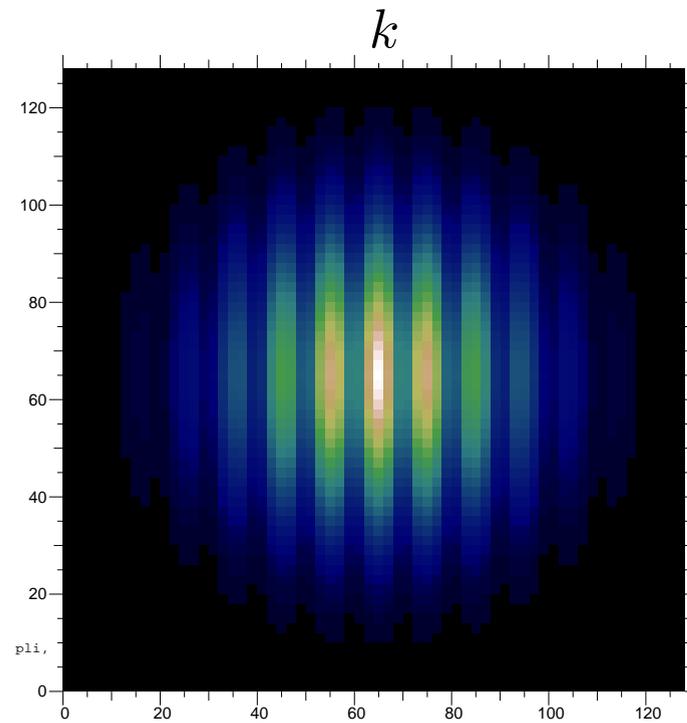
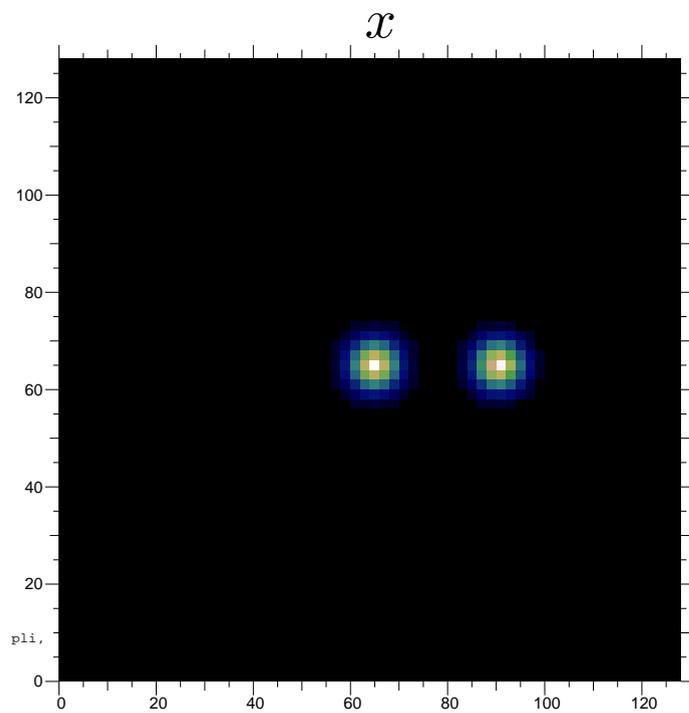
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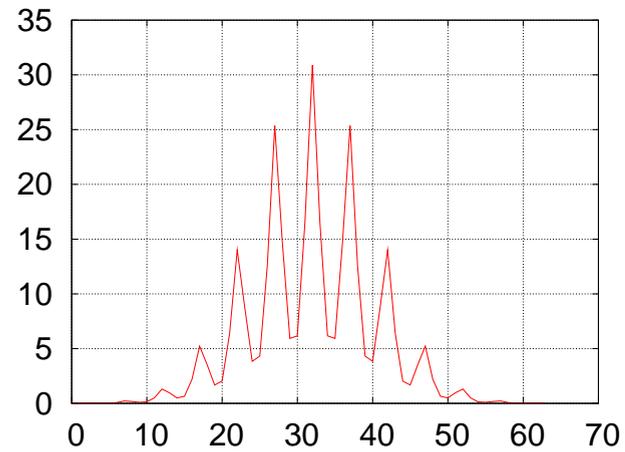
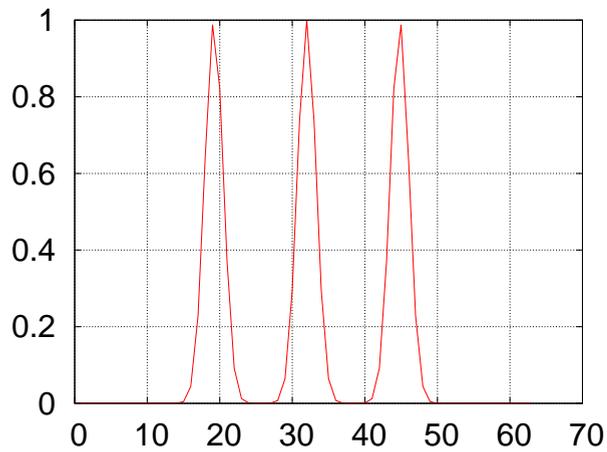
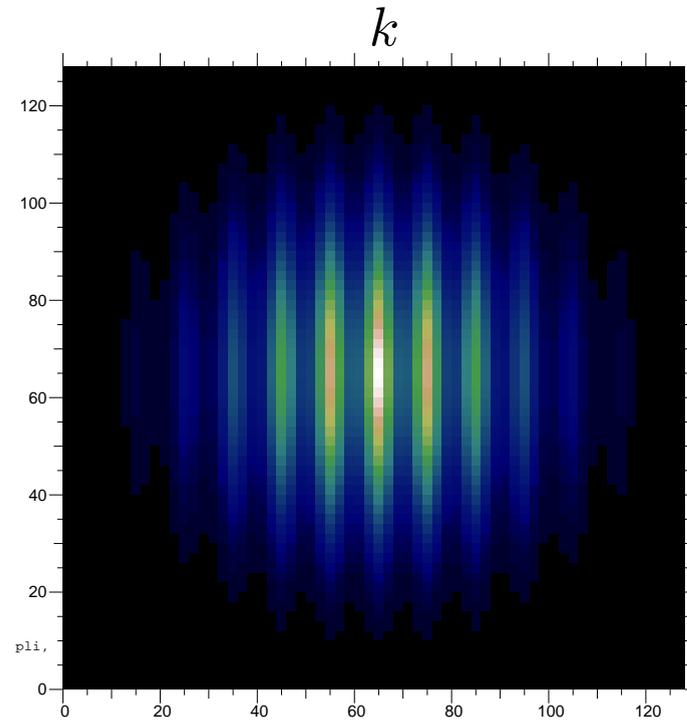
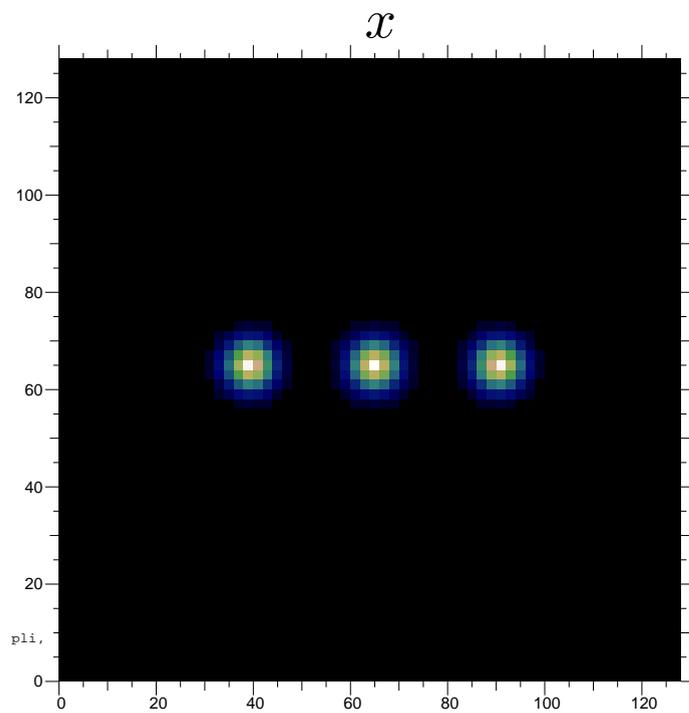
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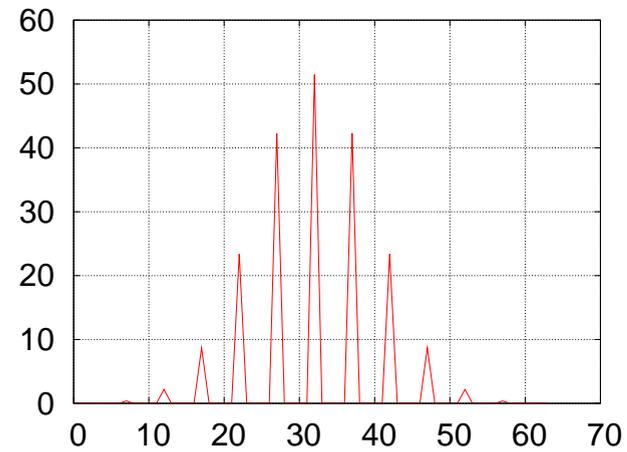
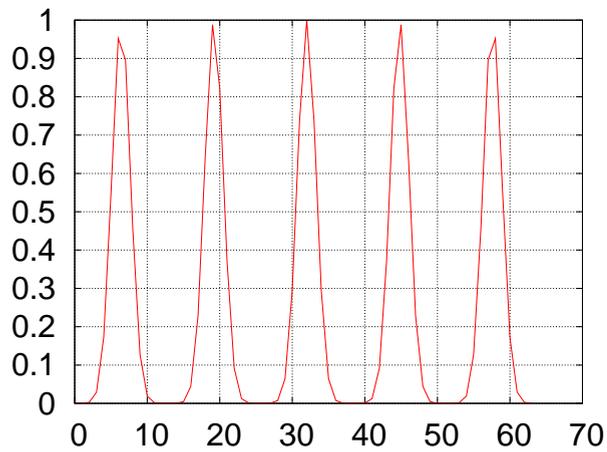
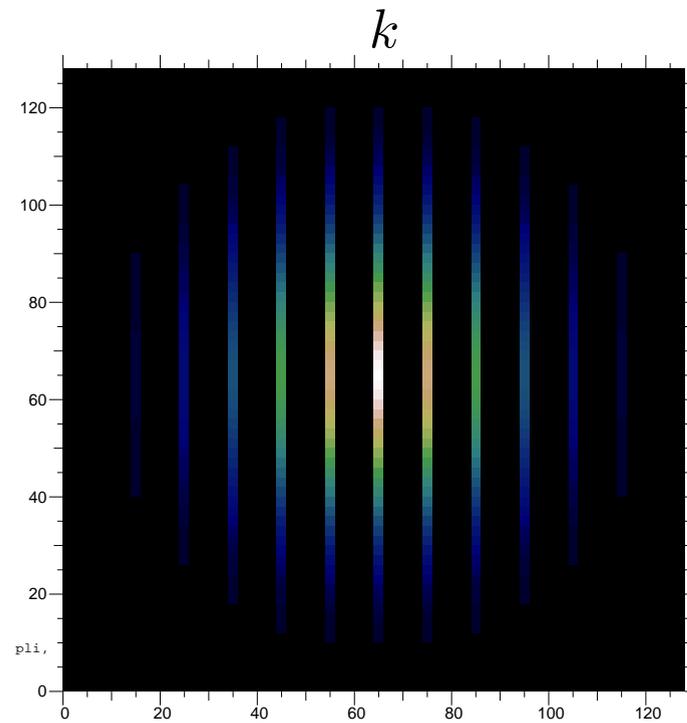
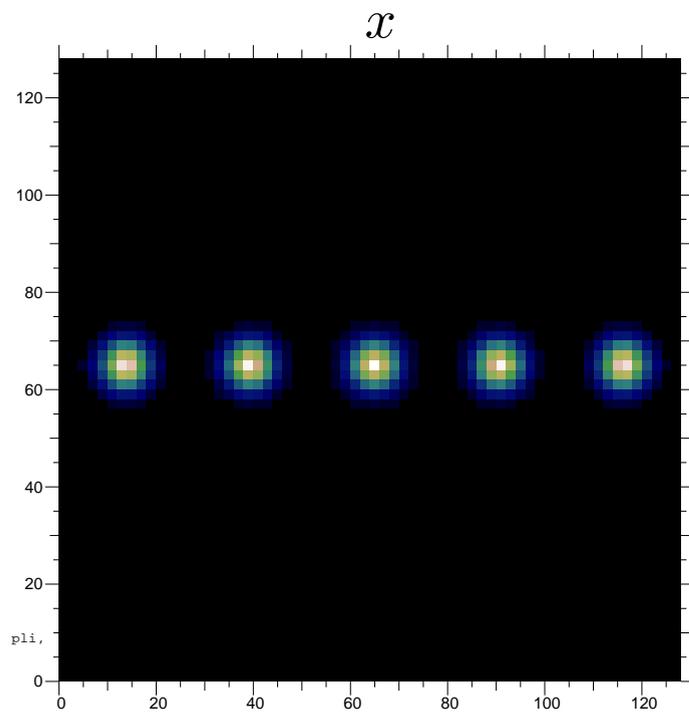
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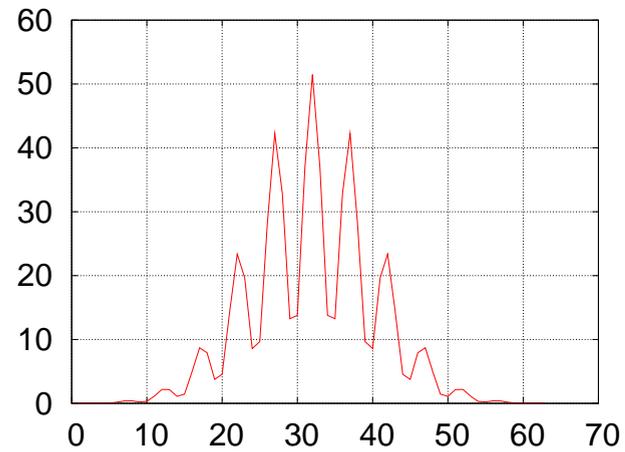
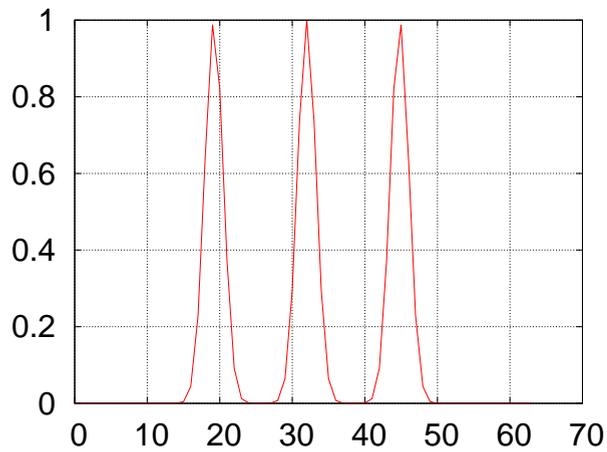
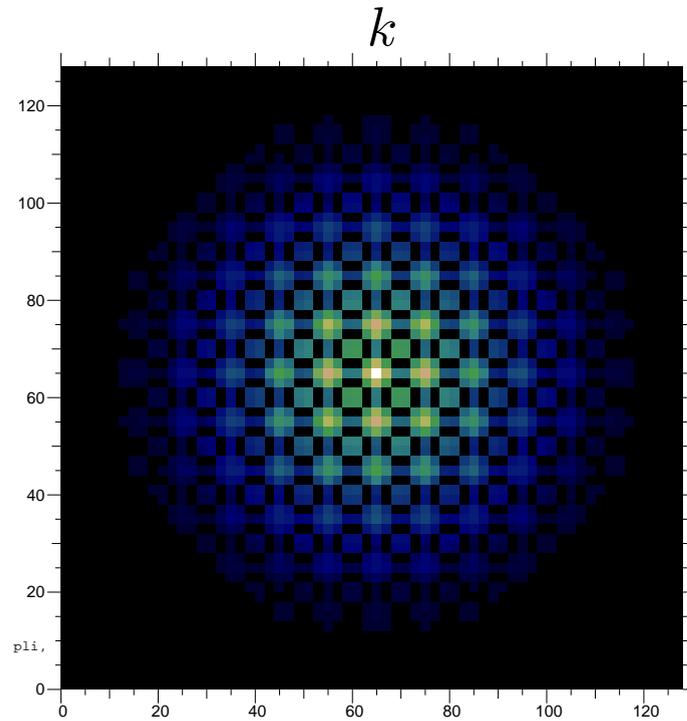
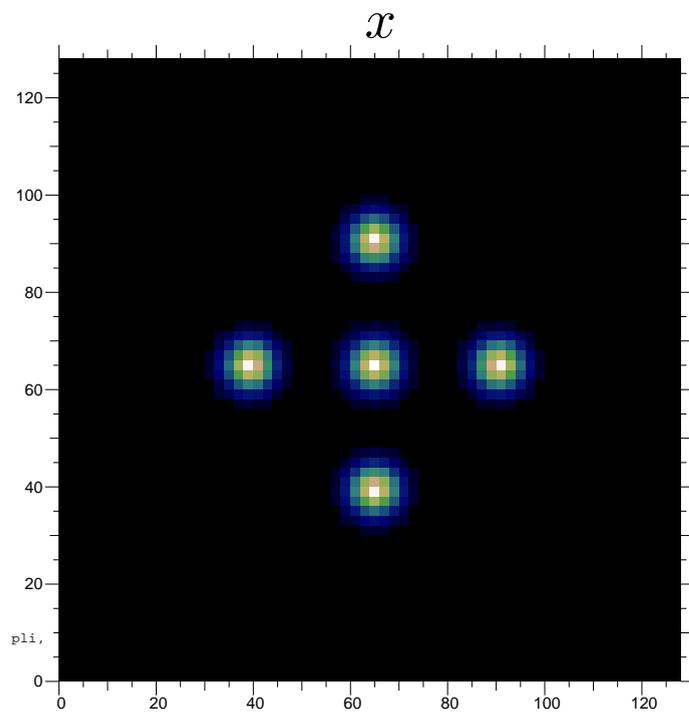
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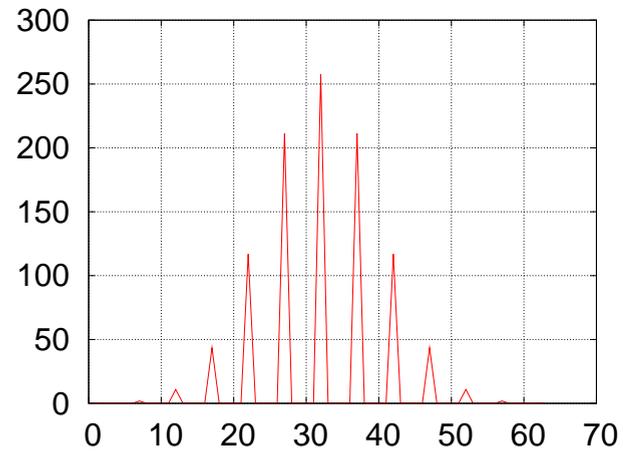
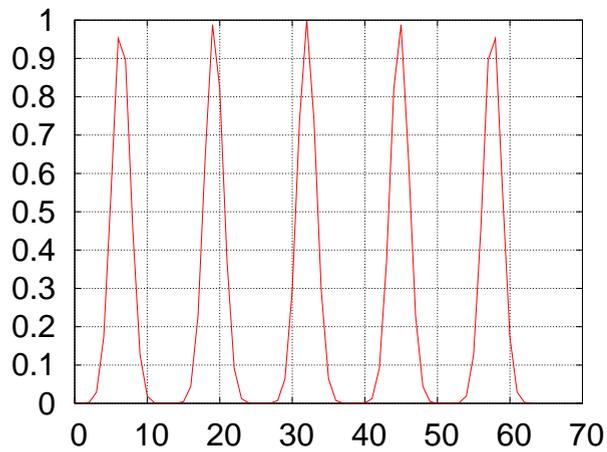
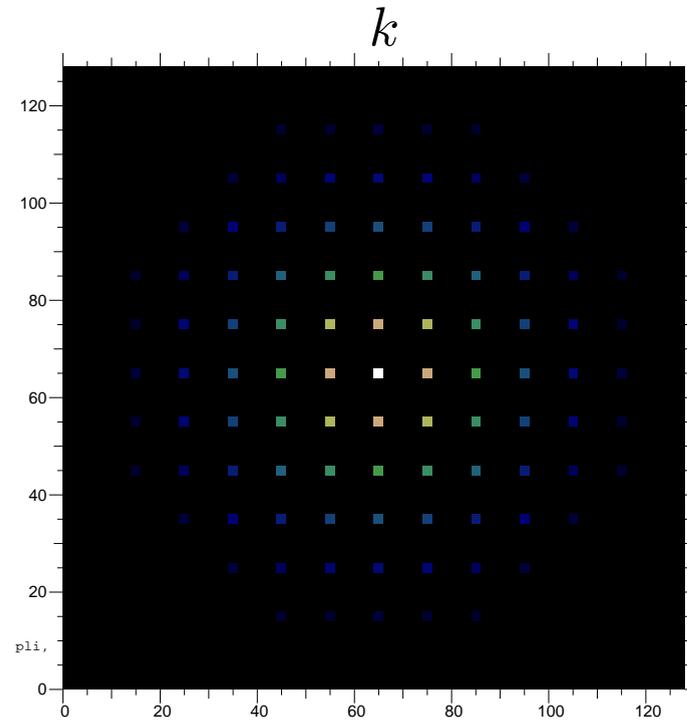
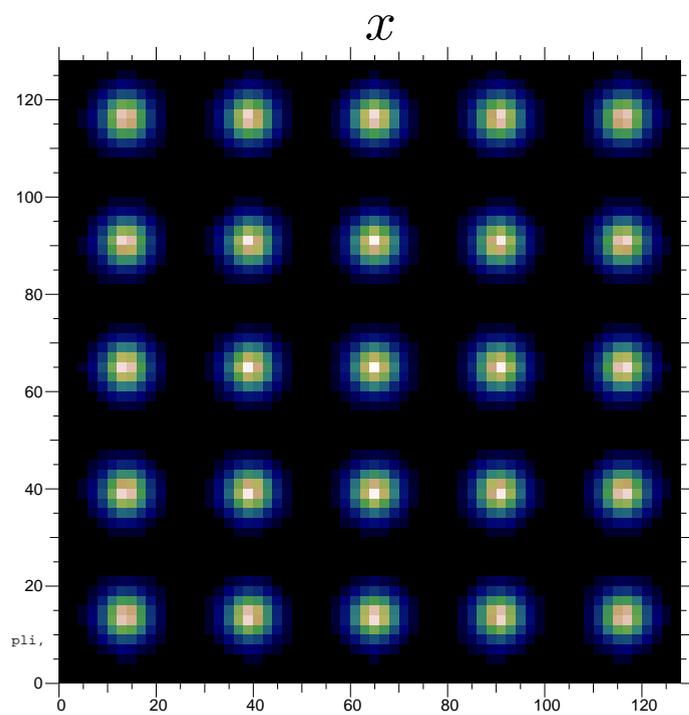
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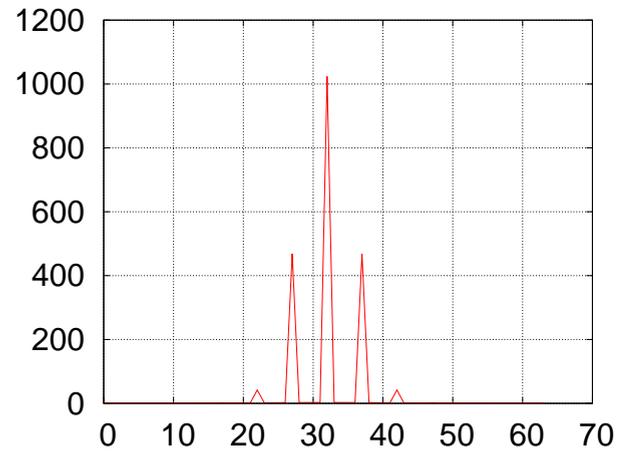
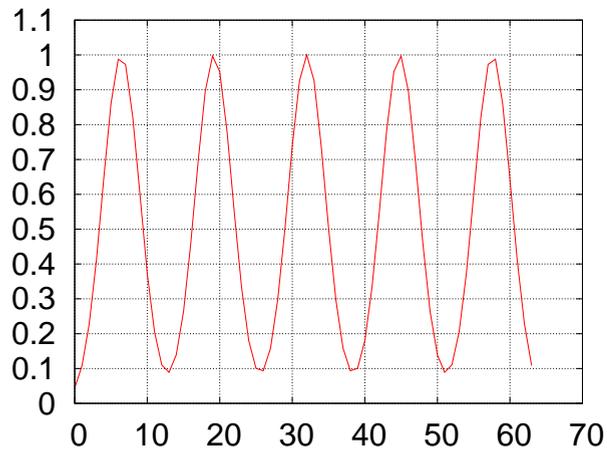
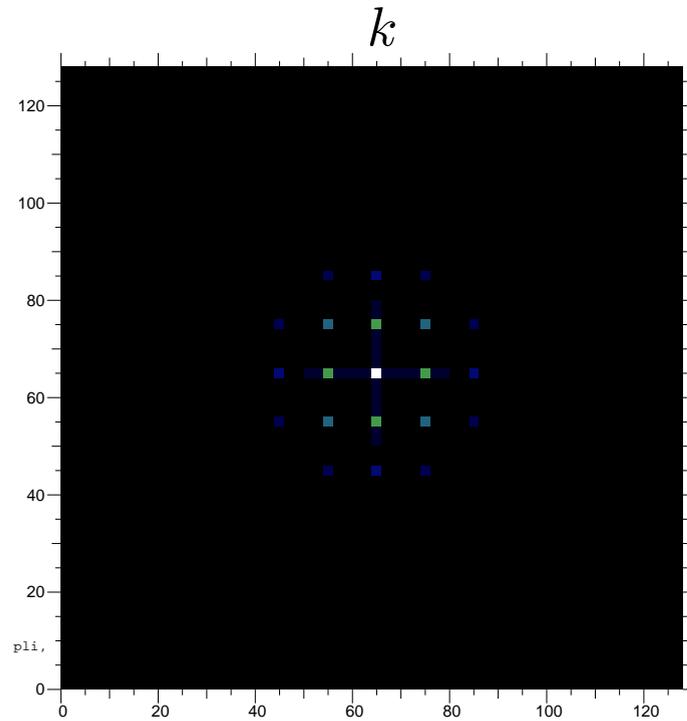
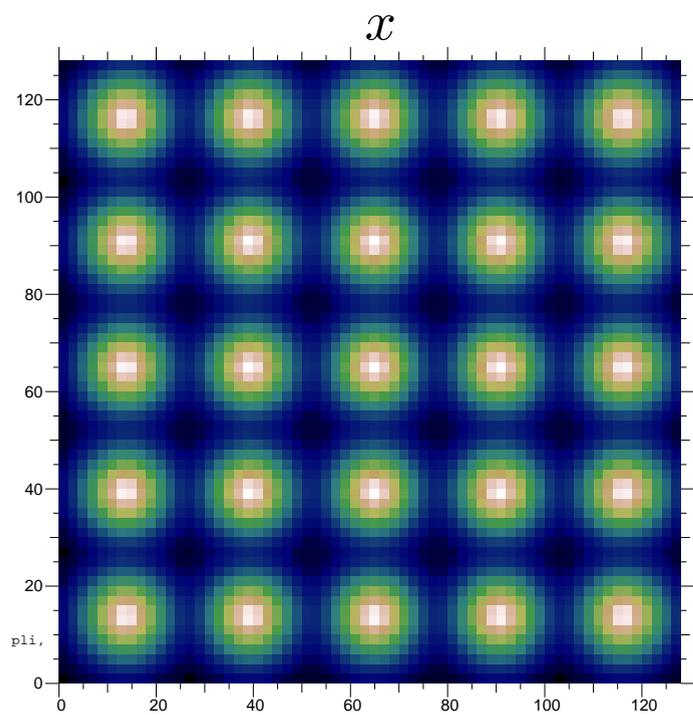
# Blobs in Fourier space



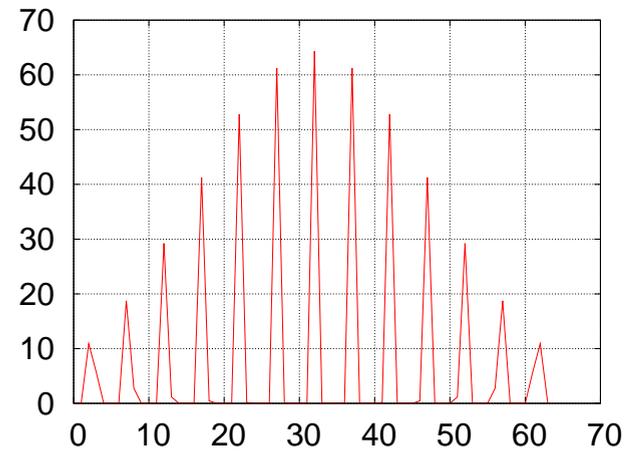
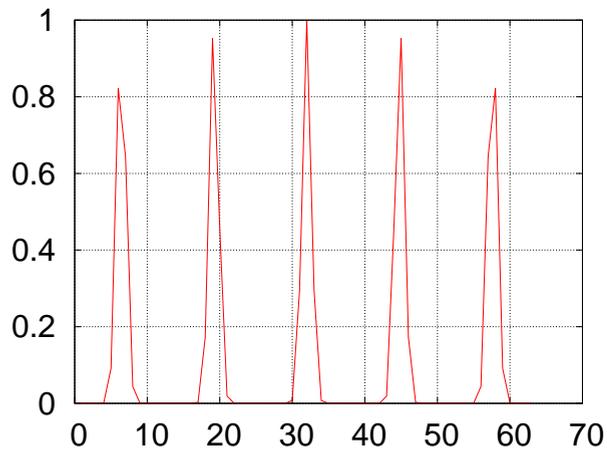
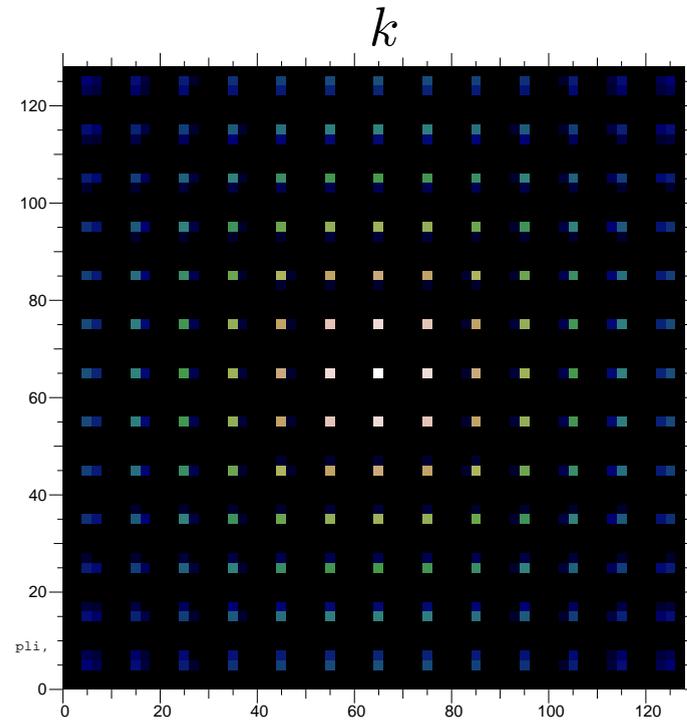
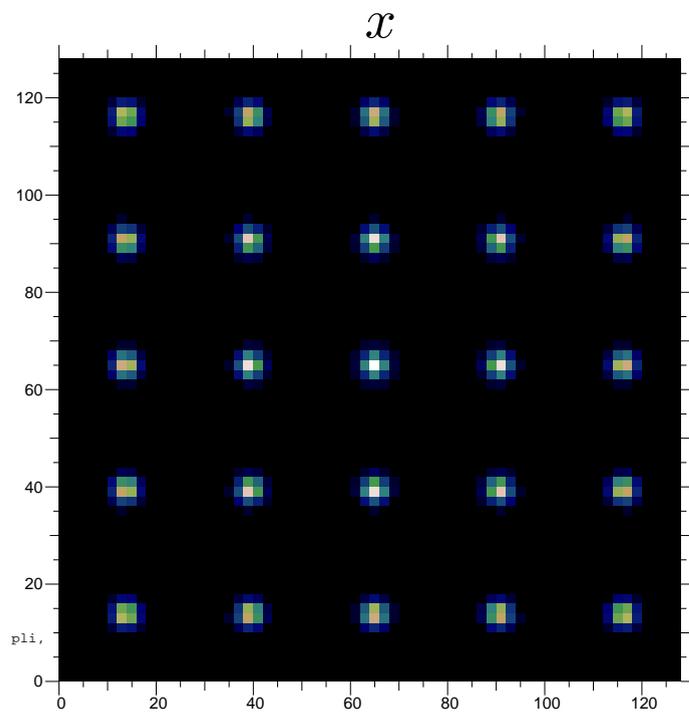
# Blobs in Fourier space



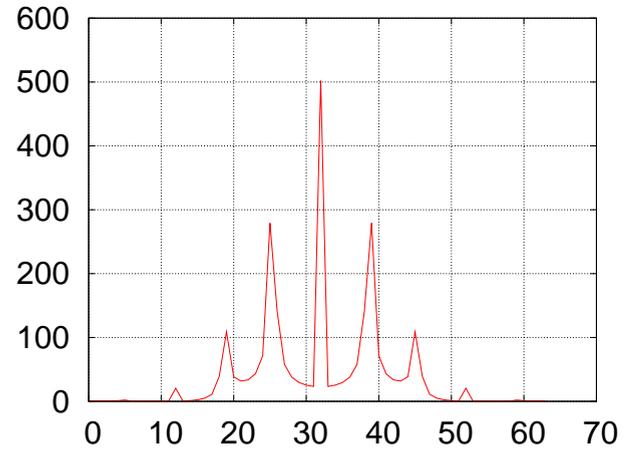
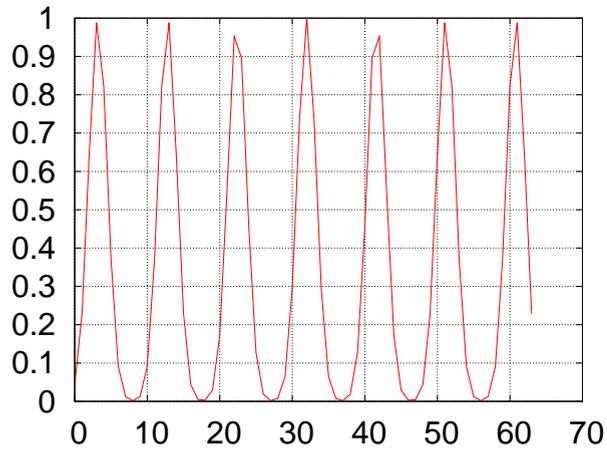
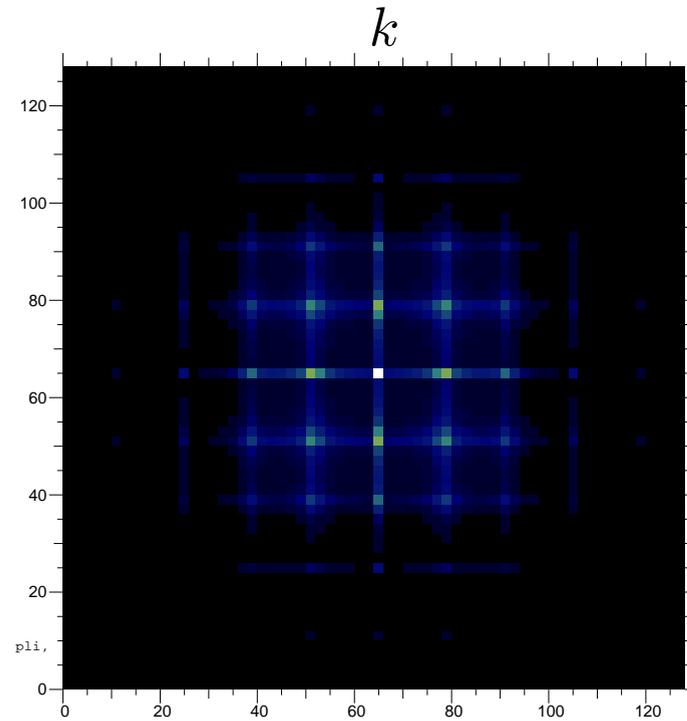
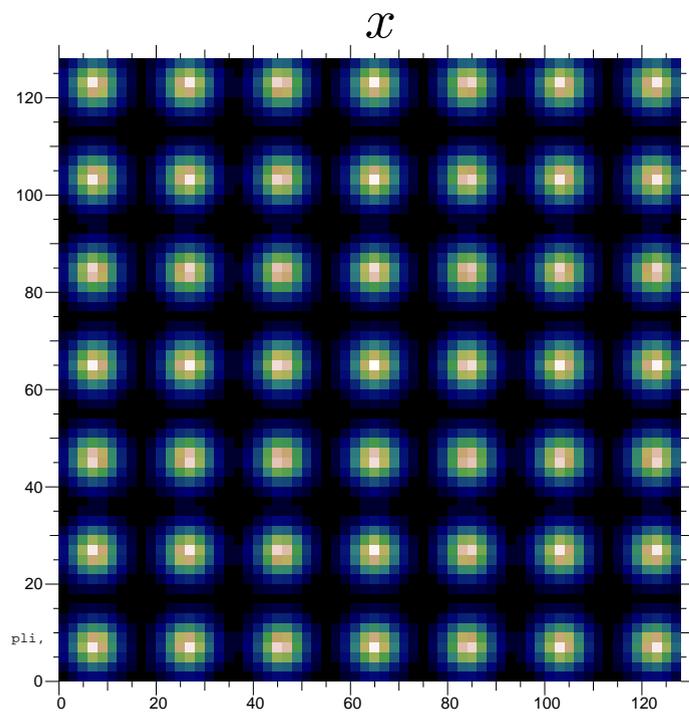
# Blobs in Fourier space



# Blobs in Fourier space



# Blobs in Fourier space



# Linear operators

General linear operator:

$$Lf(x) = \sum_y K(x, y) f(y)$$

Translation-invariant:

$$Lf(x) = \sum_y K(x - y) f(y) = \sum_y K(d) f(x - d)$$

|   |               |               |               |   |
|---|---------------|---------------|---------------|---|
| 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 |

averaging

|   |                |   |               |   |
|---|----------------|---|---------------|---|
| 0 | 0              | 0 | 0             | 0 |
| 0 | 0              | 0 | 0             | 0 |
| 0 | $-\frac{1}{2}$ | 0 | $\frac{1}{2}$ | 0 |
| 0 | 0              | 0 | 0             | 0 |
| 0 | 0              | 0 | 0             | 0 |

derivative

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

Laplacian

# Convolutions

Convolution:

$$(f * g)(x) = \sum_y f^*(x - y)g(y) = \sum_y f^*(y)g(x - y)$$

Template matching:

- $g(x)$  is template function, centered around  $x = 0$
- $(f * g)(x)$  gives *overlap* of  $f$ , when  $g$  is  $f$  is translated to  $x$

Smoothing:

- $g(x)$  is smoothing kernel, centered around  $x = 0$
- $(f * g)(x)$  is  $f(x)$  with each value replaced by a local average

# Convolutions in Fourier space

$$\begin{aligned}\mathcal{F}(f * g)(k) &= \sum_x (f * g)(x) e^{-ik \cdot x} \\ &= \sum_x \sum_y f(x - y) g^*(y) e^{-ik \cdot x} \\ &= \sum_x \sum_y \sum_{k_1, k_2} \hat{f}(k_1) e^{ik_1 \cdot (x - y)} \hat{g}^*(k_2) e^{-ik_2 \cdot y} e^{-ik \cdot x} \\ &= \sum_{k_1, k_2} \hat{f}(k_1) \hat{g}^*(k_2) \sum_{x, y} e^{i(k_1 - k) \cdot x} e^{-i(k_2 - k) \cdot y} \\ &= \hat{f}^*(-k) \hat{g}(-k)\end{aligned}$$

using orthogonality:

$$\sum_x e^{ik \cdot (x - y)} = \delta_{x, y}$$

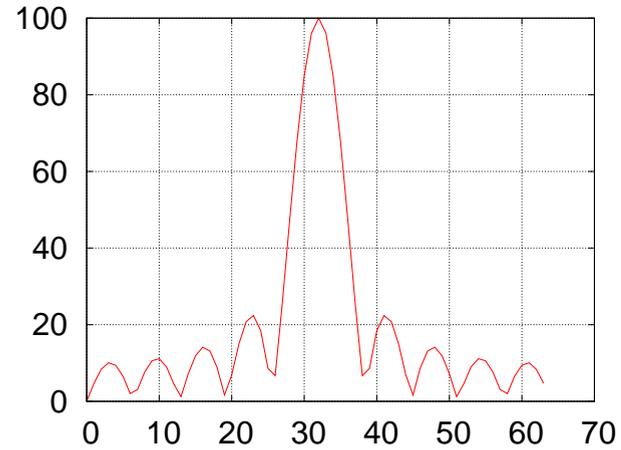
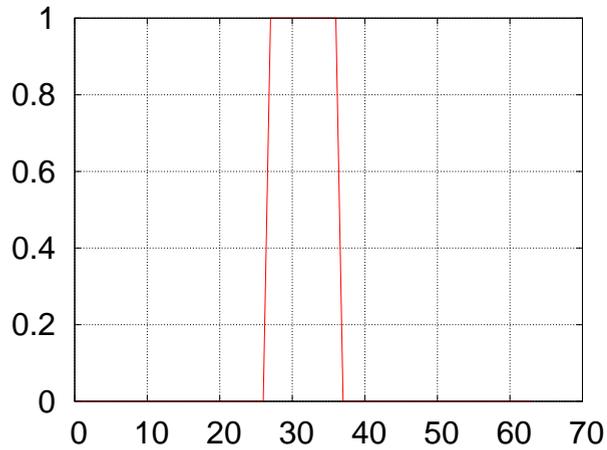
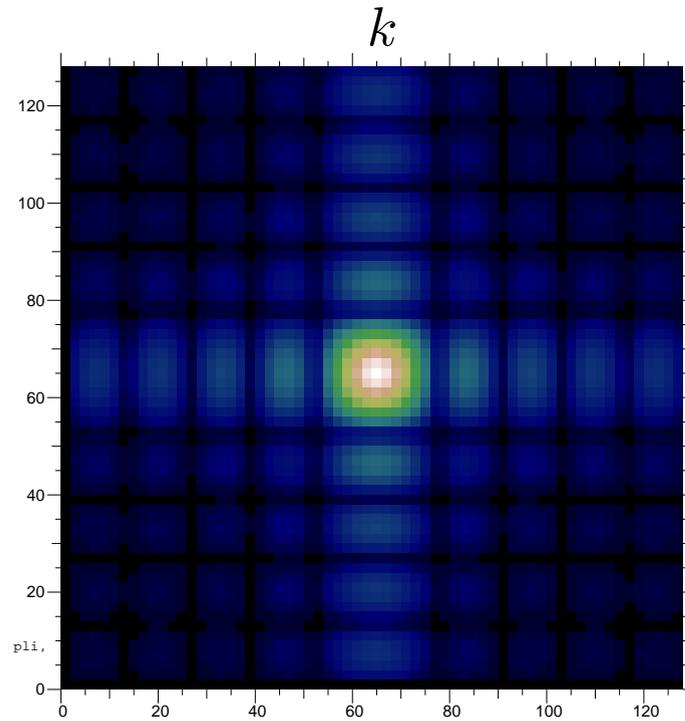
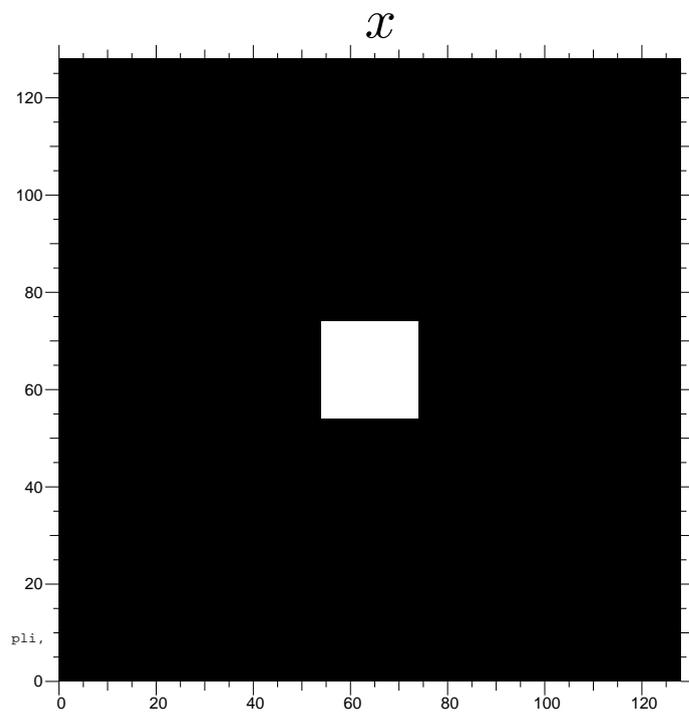
# Convolutions in Fourier space

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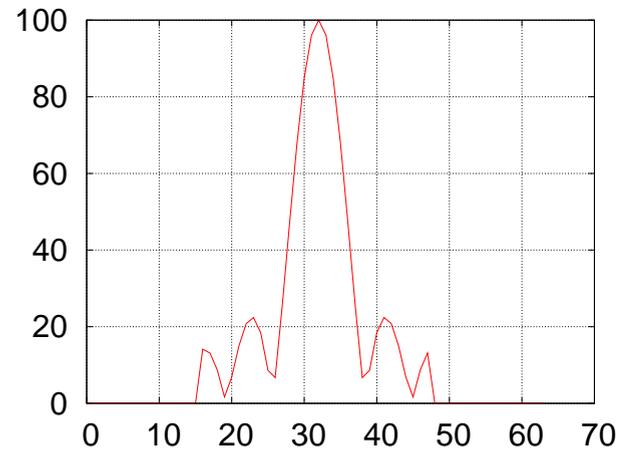
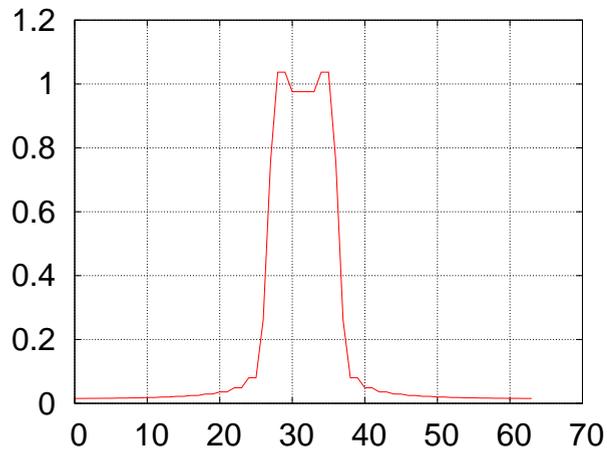
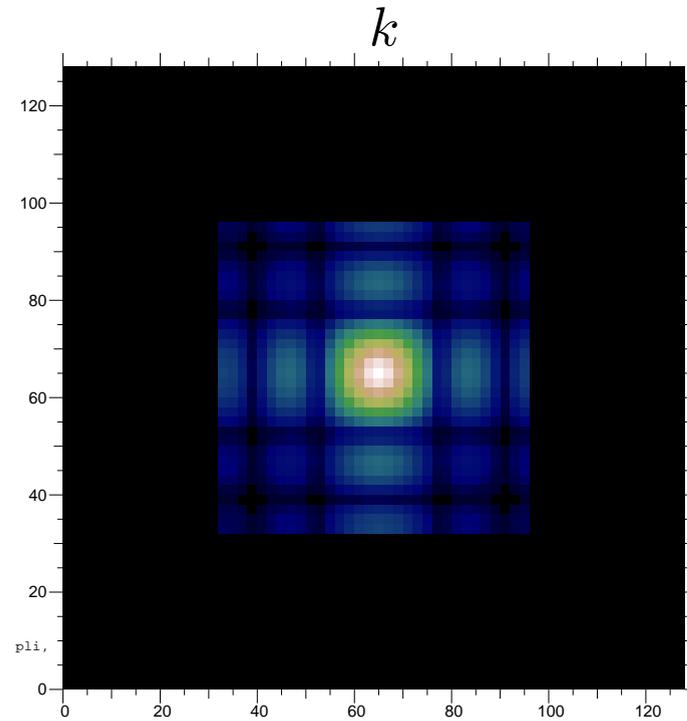
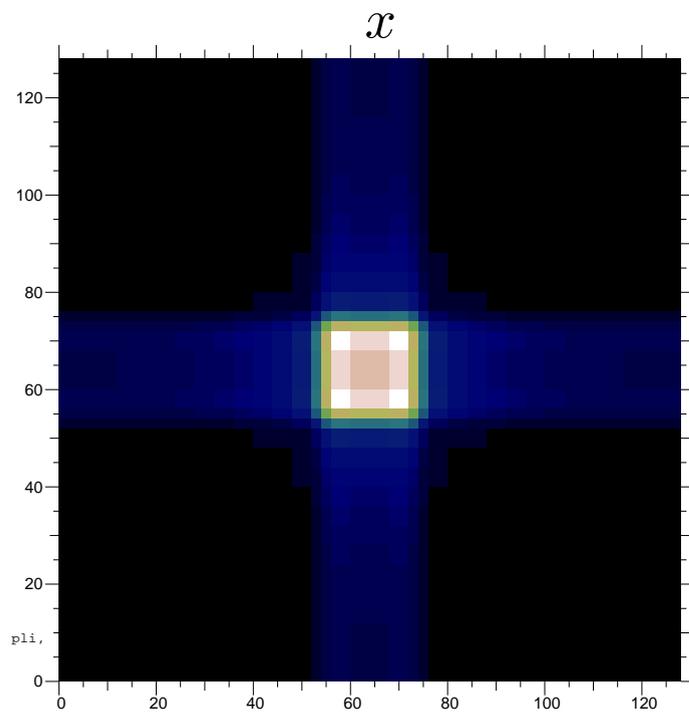
$$\mathcal{F}(f * g)(k) = \hat{f}^*(-k) \hat{g}(-k)$$

- Convolution in real space is multiplication in Fourier space
- Fast algorithm for
  - Filtering
  - Matching
  - Numerical operations

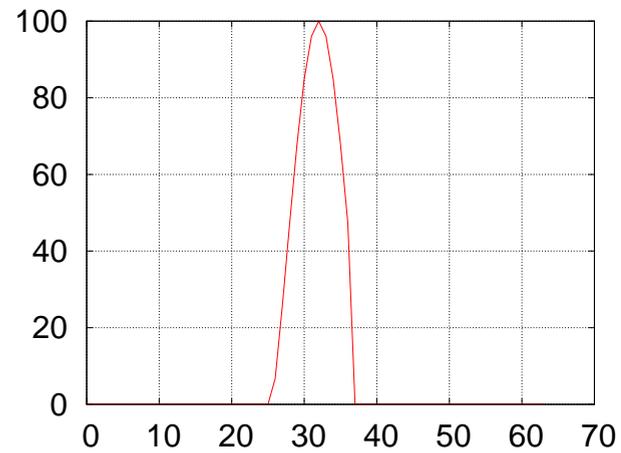
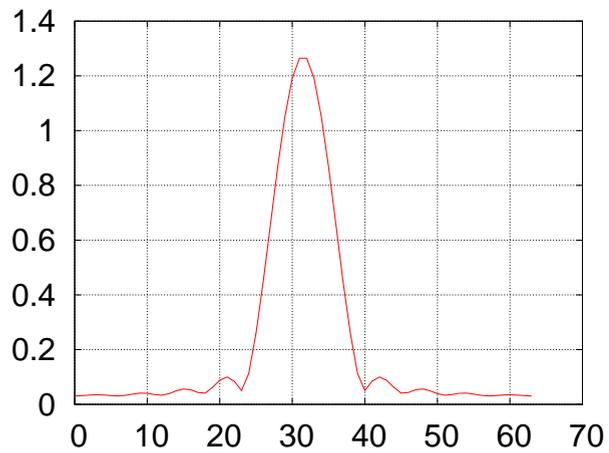
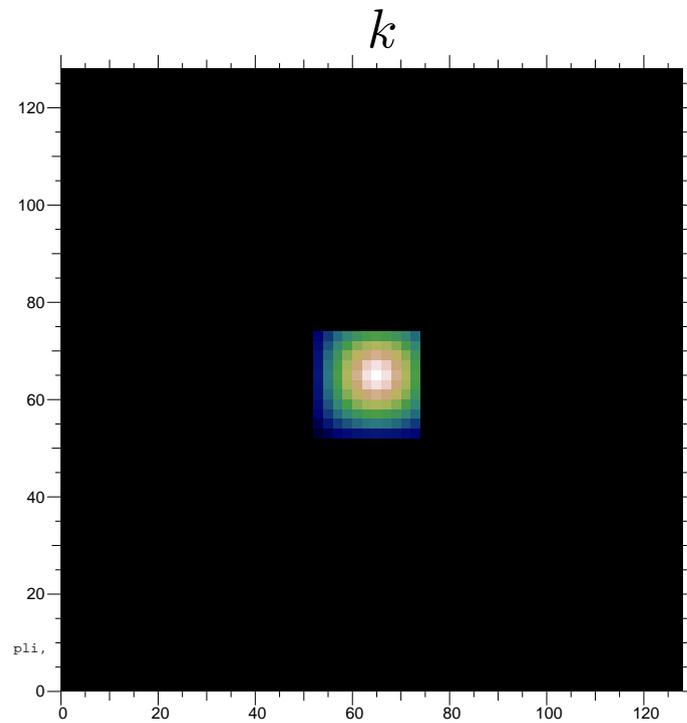
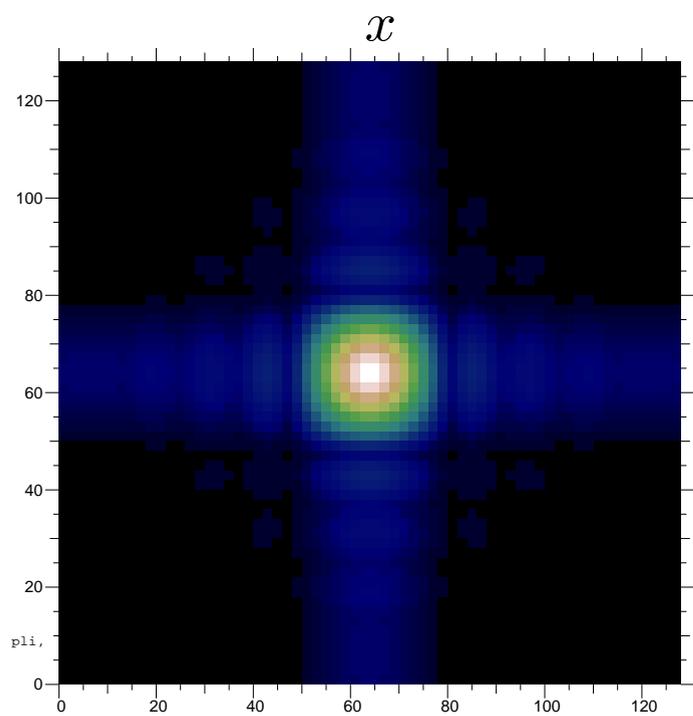
# Convolution in Fourier space



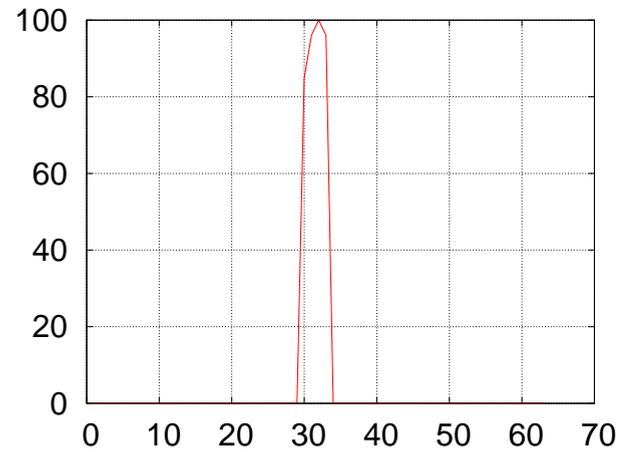
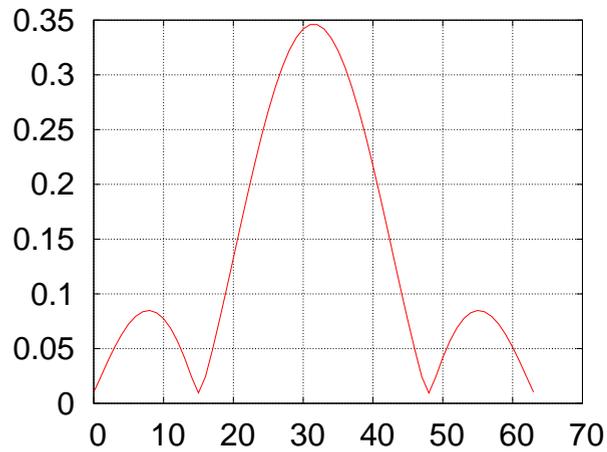
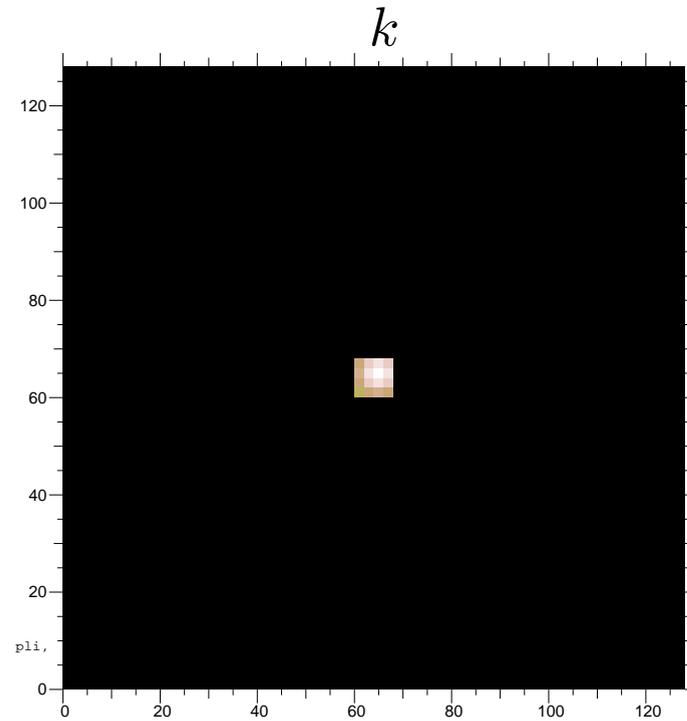
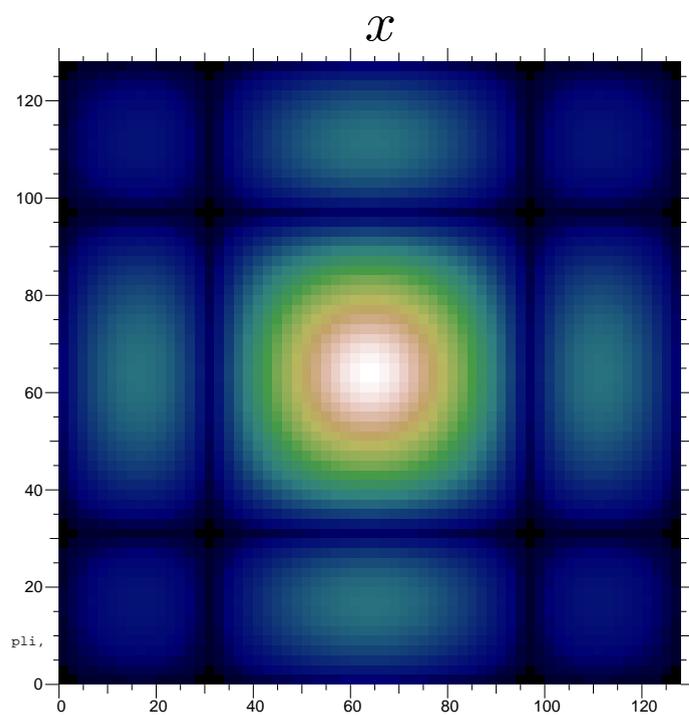
# Convolution in Fourier space



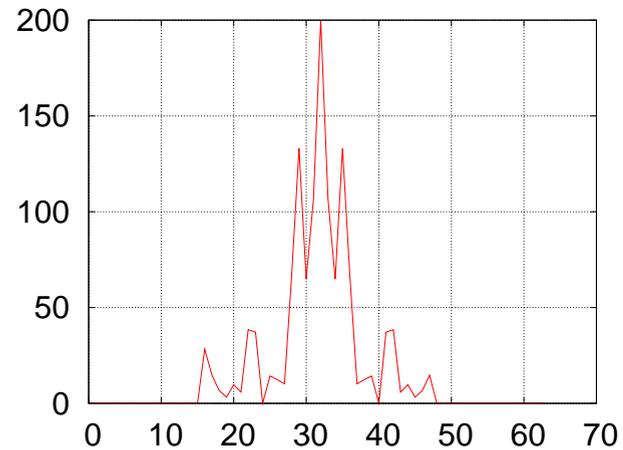
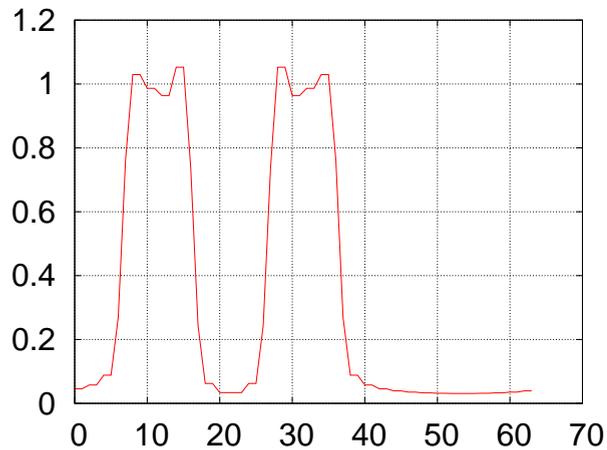
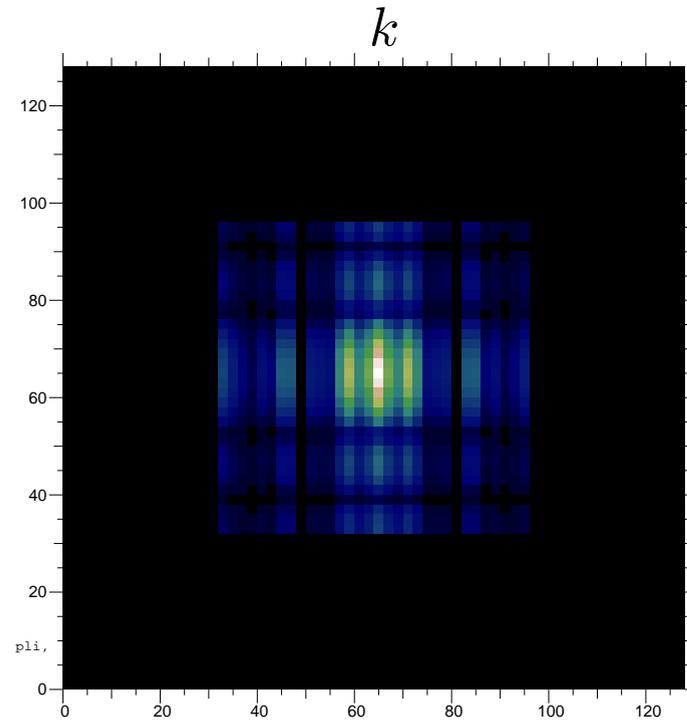
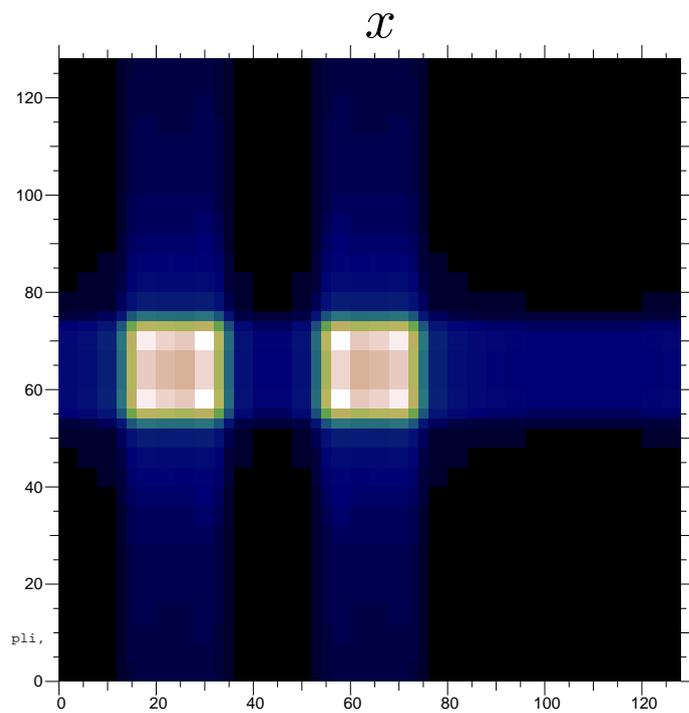
# Convolution in Fourier space



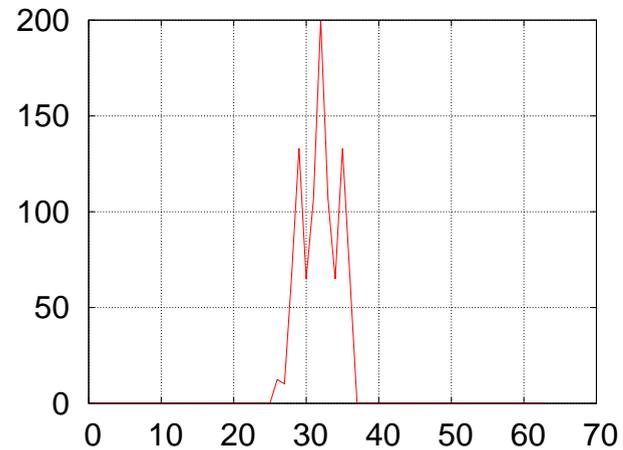
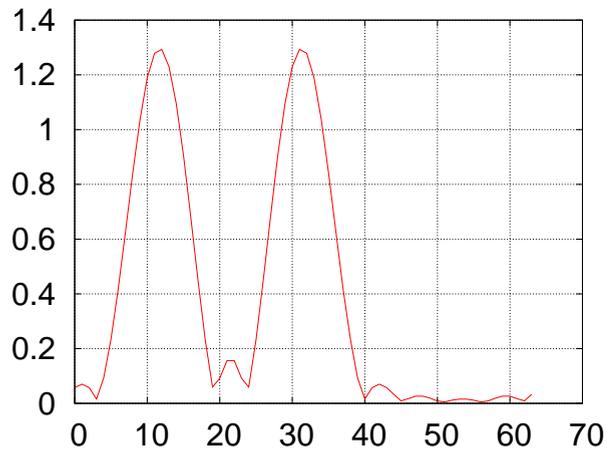
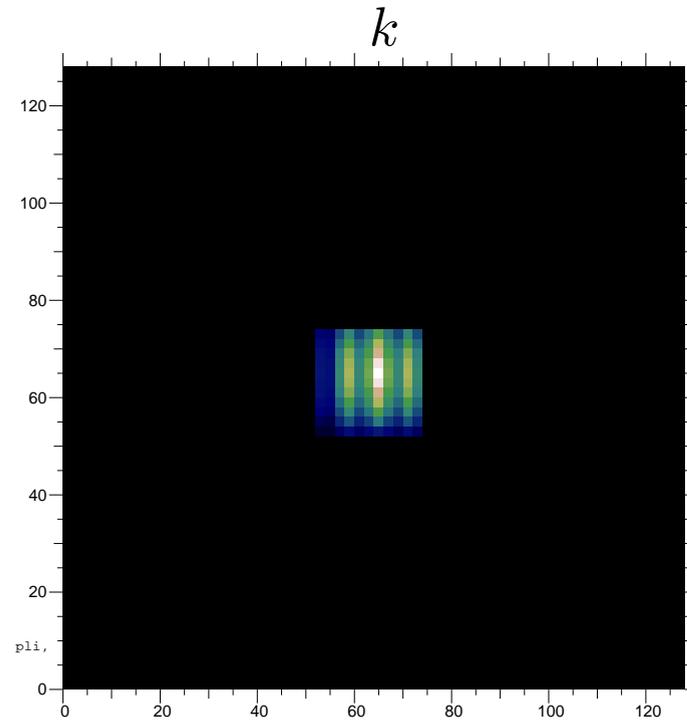
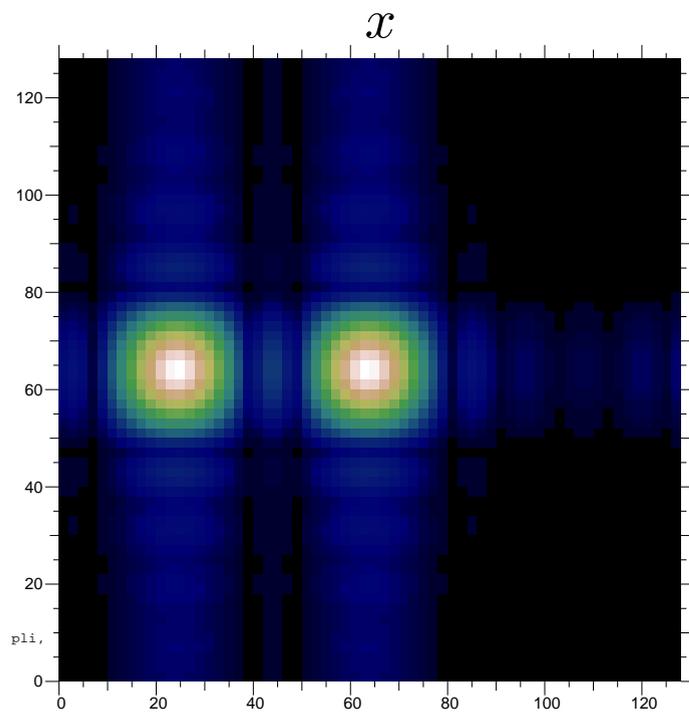
# Convolution in Fourier space



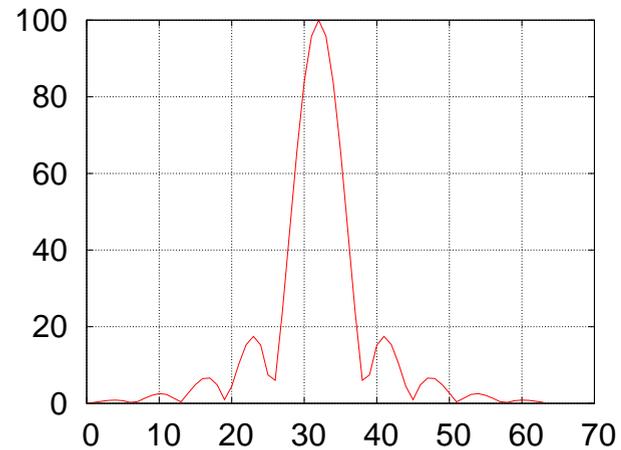
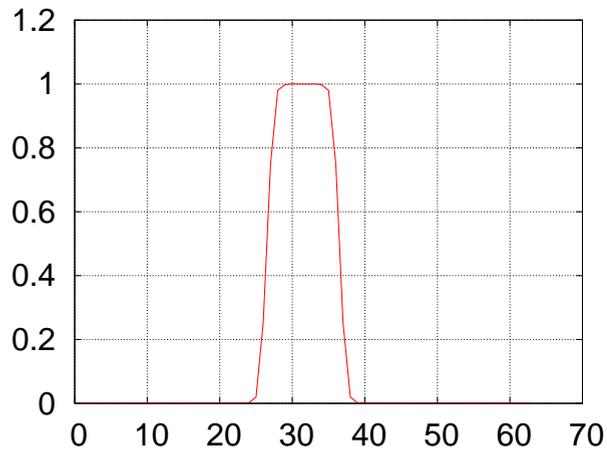
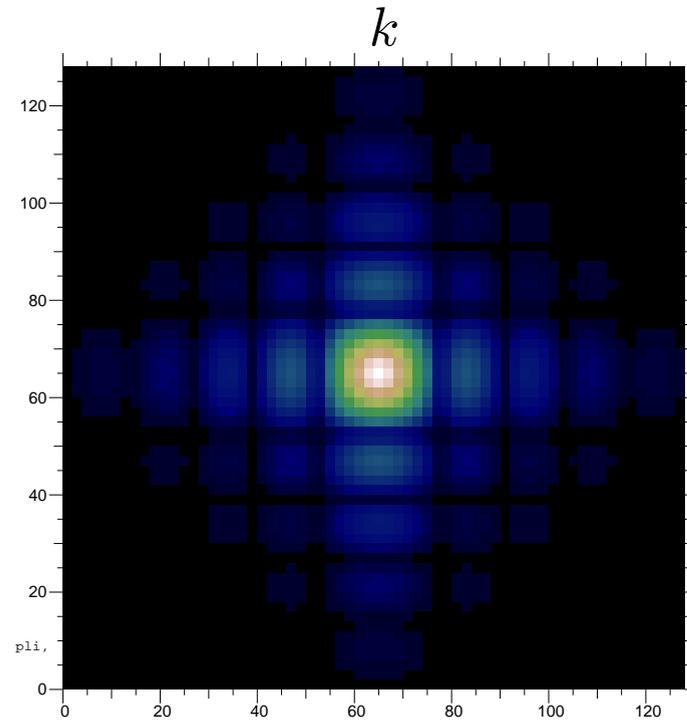
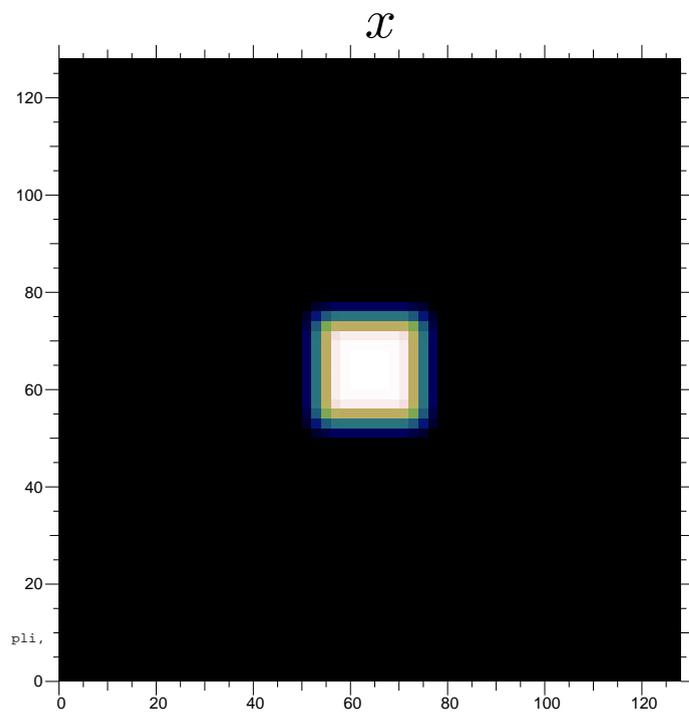
# Convolution in Fourier space



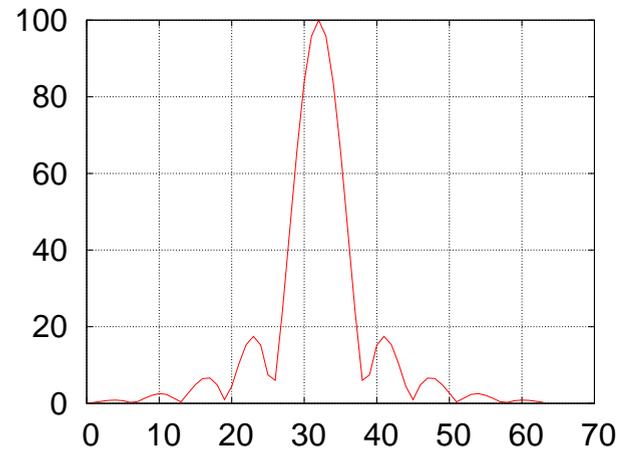
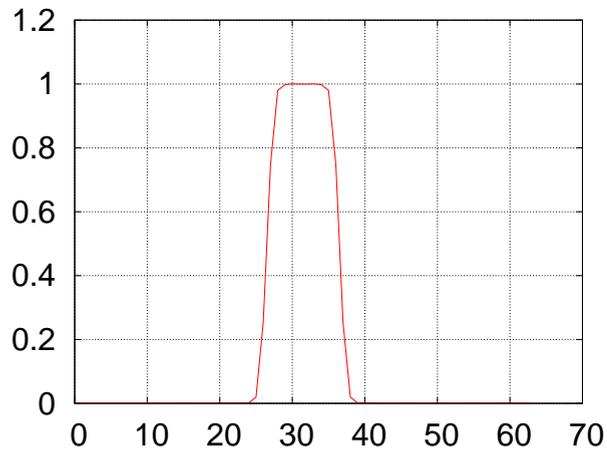
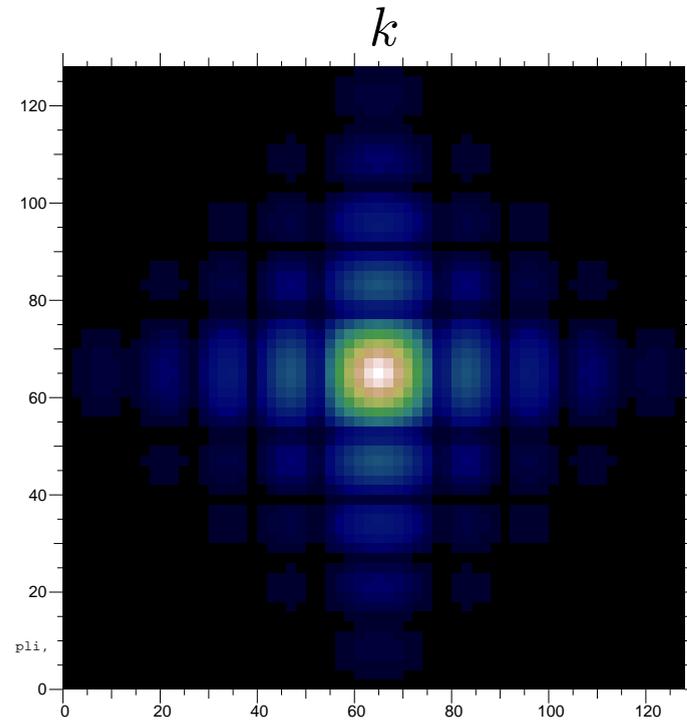
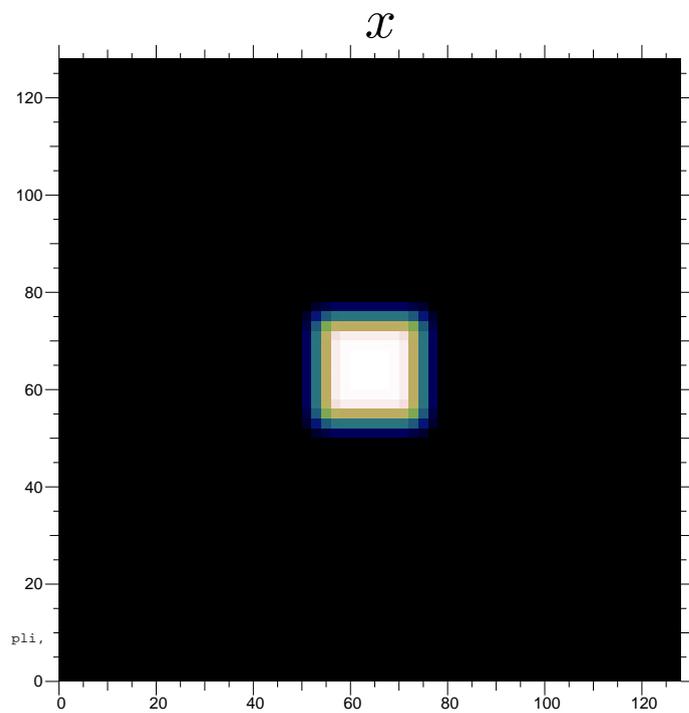
# Convolution in Fourier space



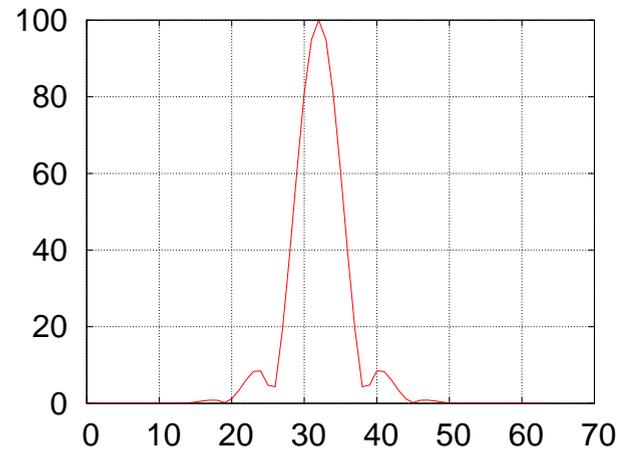
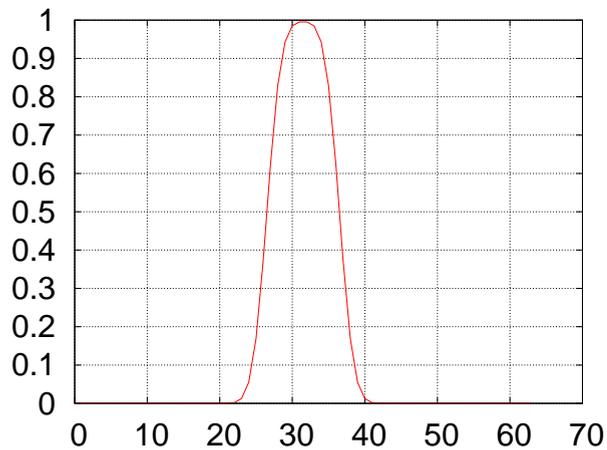
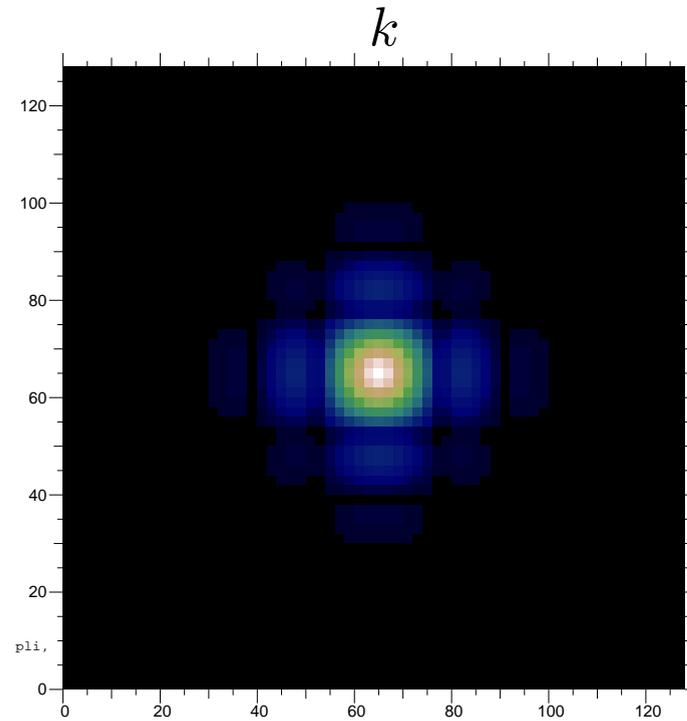
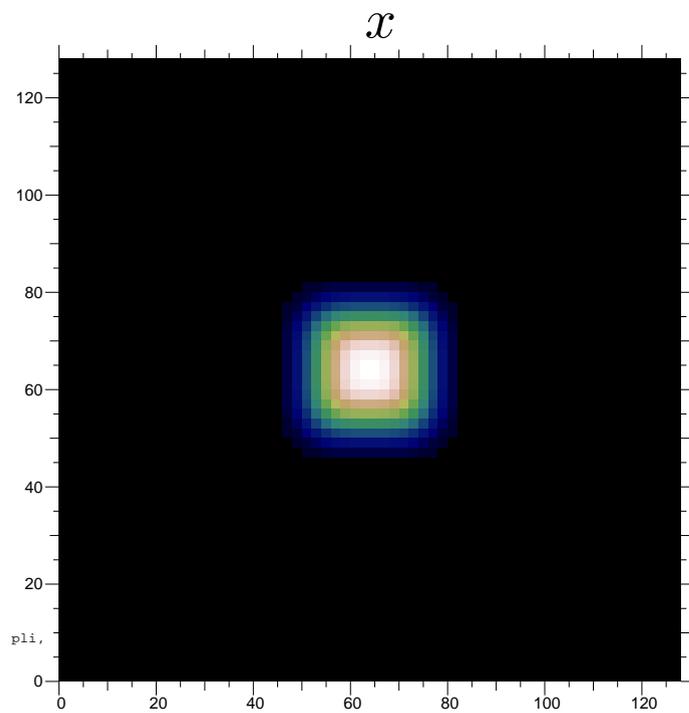
# Convolution in Fourier space



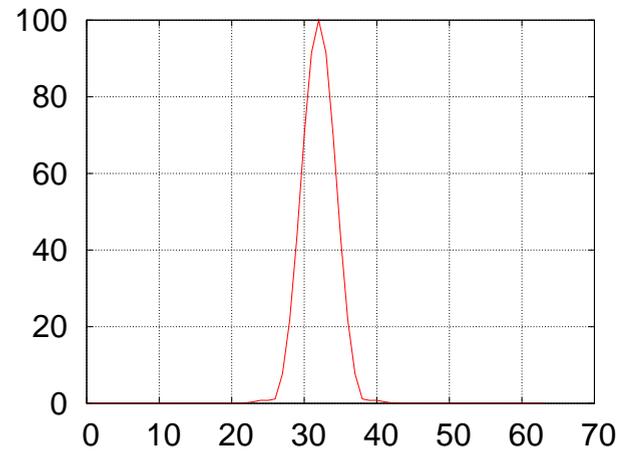
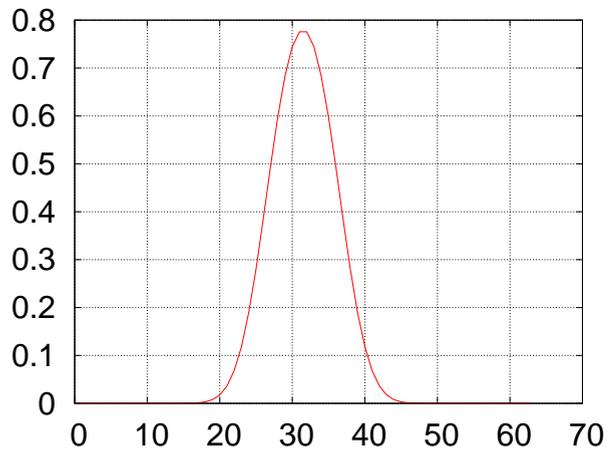
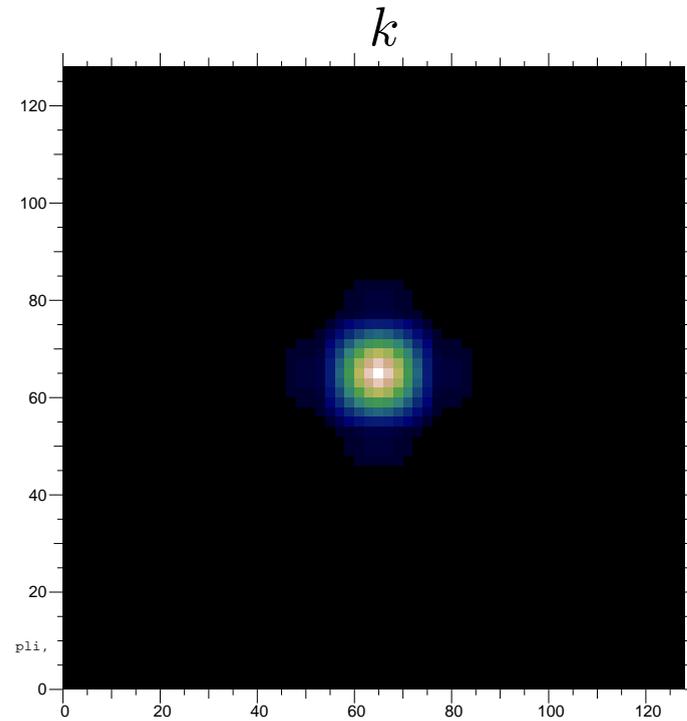
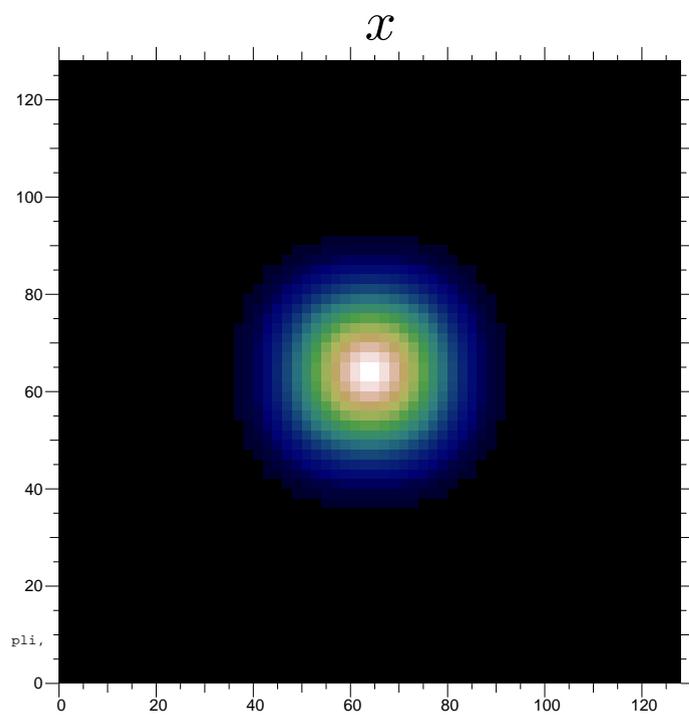
# Convolution in Fourier space



# Convolution in Fourier space



# Convolution in Fourier space



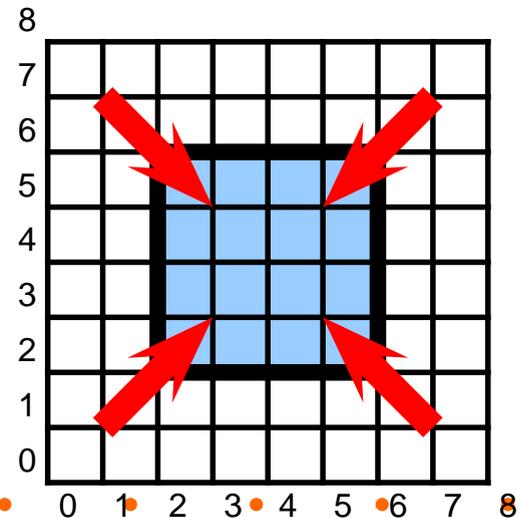
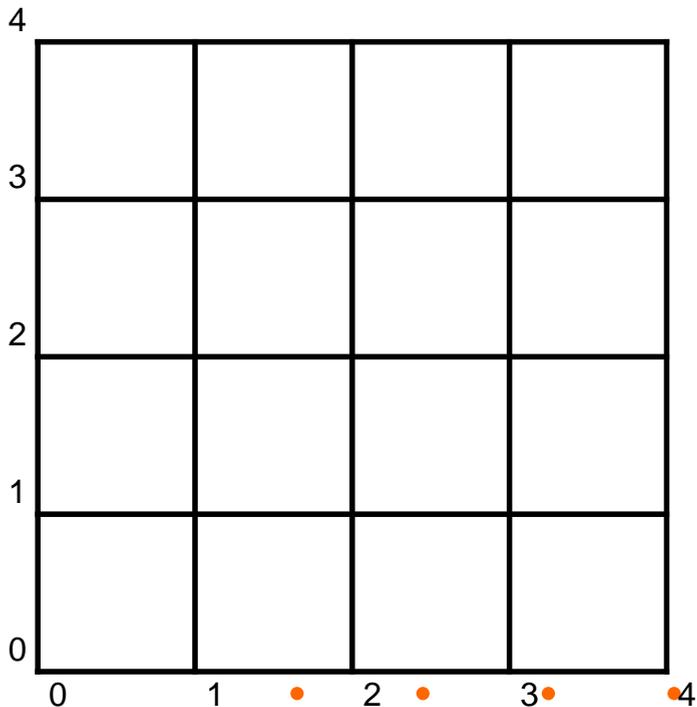
# Aliasing

Resampling: Leave out every other pixel

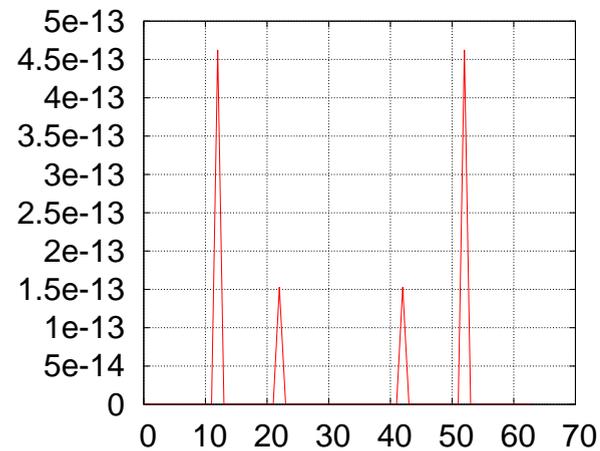
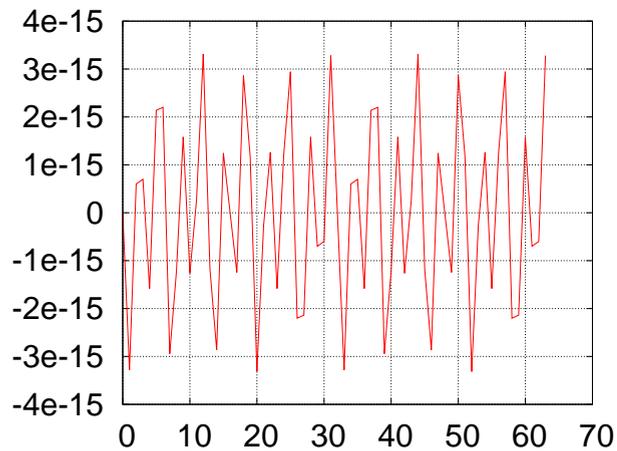
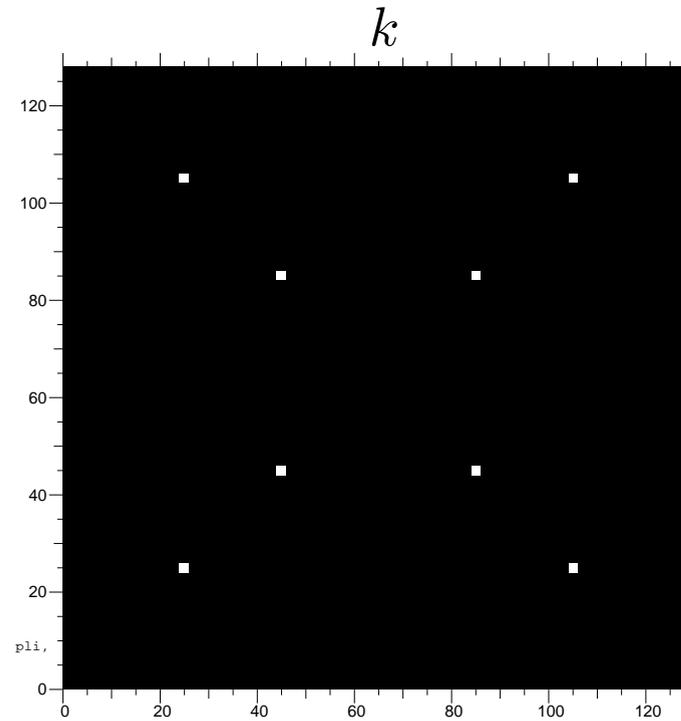
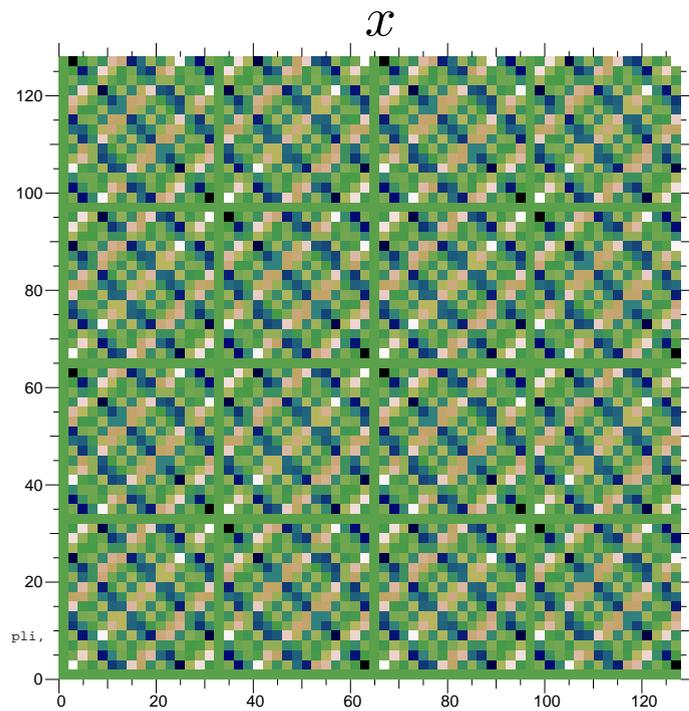
$$\Delta x \longrightarrow 2\Delta x \quad k_{\max} \longrightarrow \frac{k_{\max}}{2}$$

In Fourier space

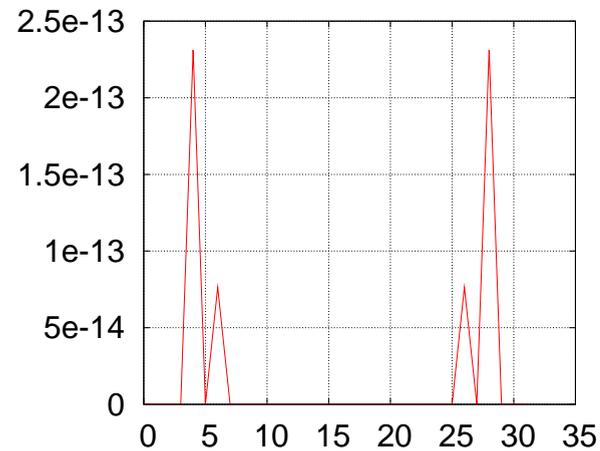
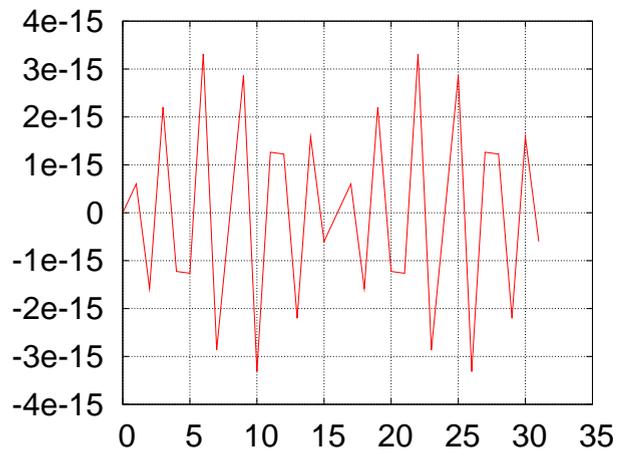
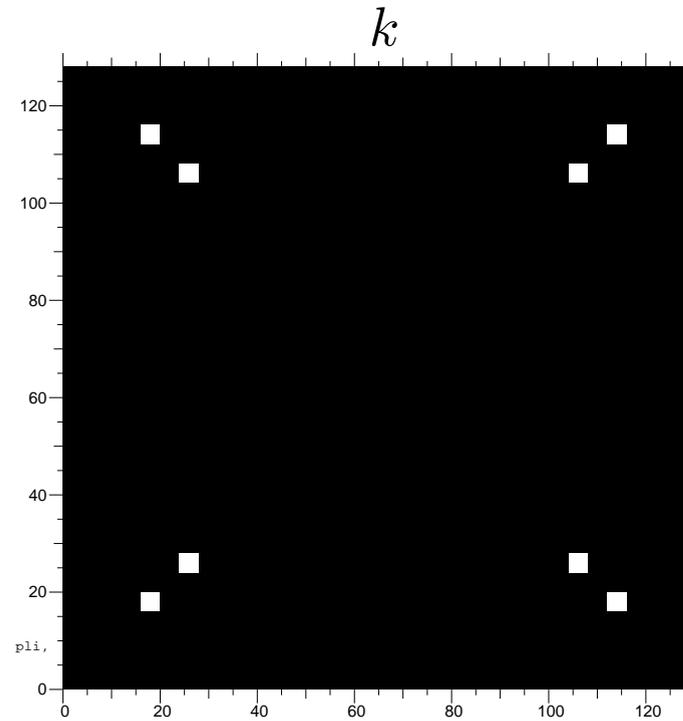
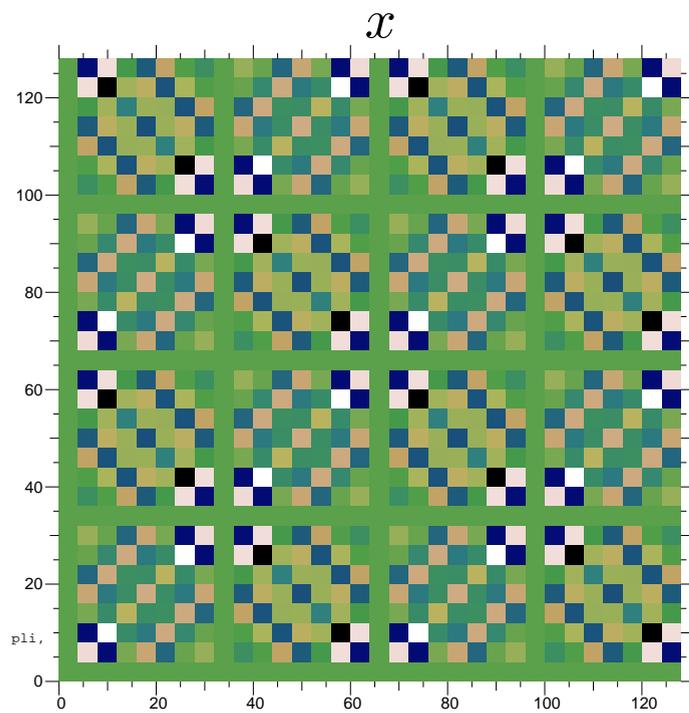
$$\hat{g}(k) = \hat{f}(k') + \hat{f}\left(k' + \frac{\pi}{\Delta x}\right)$$



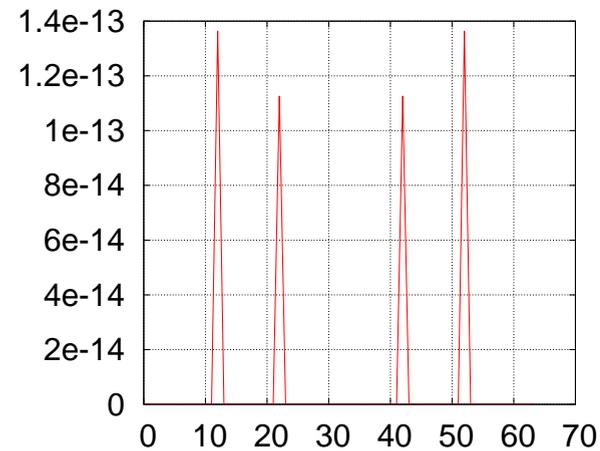
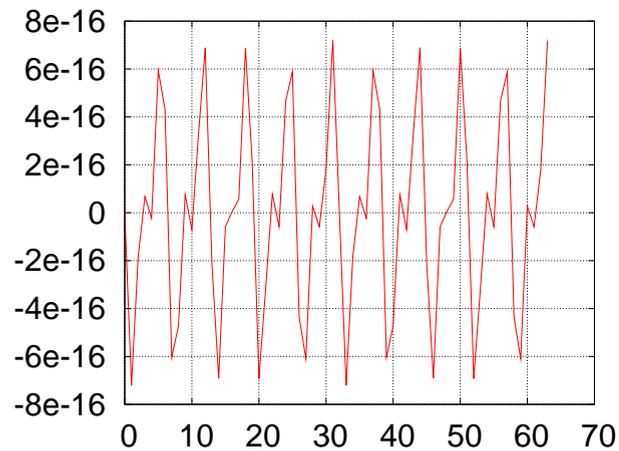
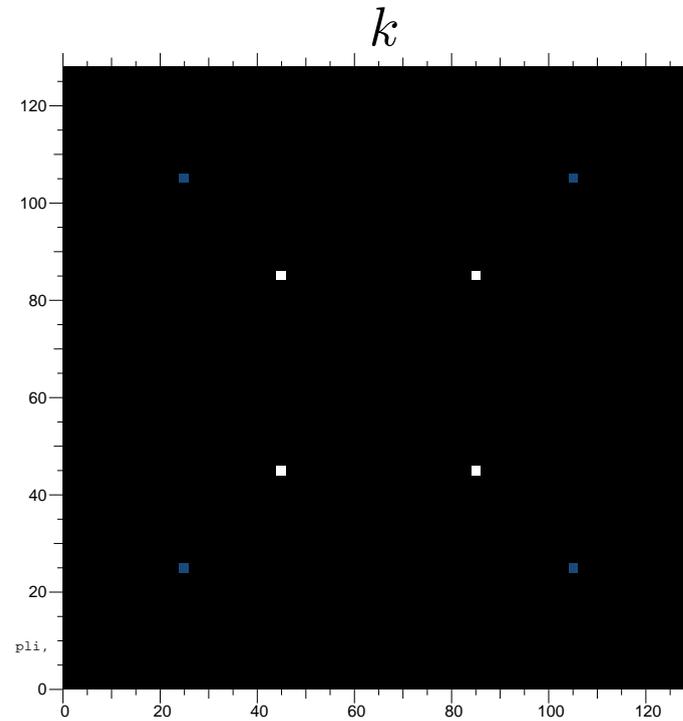
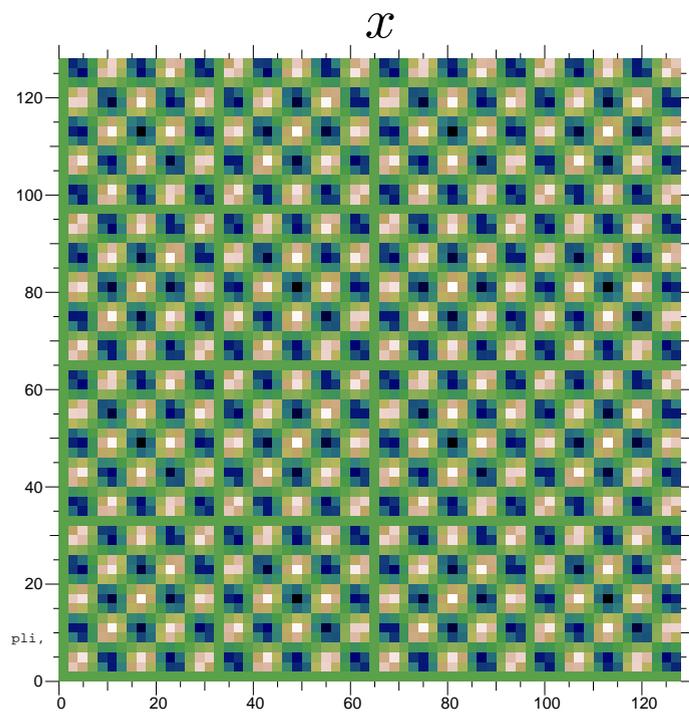
# Aliasing: high-resolution



# Aliasing: downsampled



# Aliasing: filtered



# Aliasing: filtered and downsampled

