



EECE 4353 Image Processing

Lecture Notes: Sampling and Aliasing

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Warning:

Image Resampling Can Lead to *the Jaggies!*



The jaggies!





This is a bad way to do it.

Downsampling (Decimation)



```
>> J = I(1:2:R, 1:2:C, :);
```



E.g.: every 2nd pixel in every 2nd row

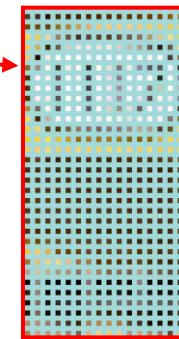


Bad, bad – very bad.

Downsampling (Decimation)



```
>> J = I(1:2:R, 1:2:C, :);
```

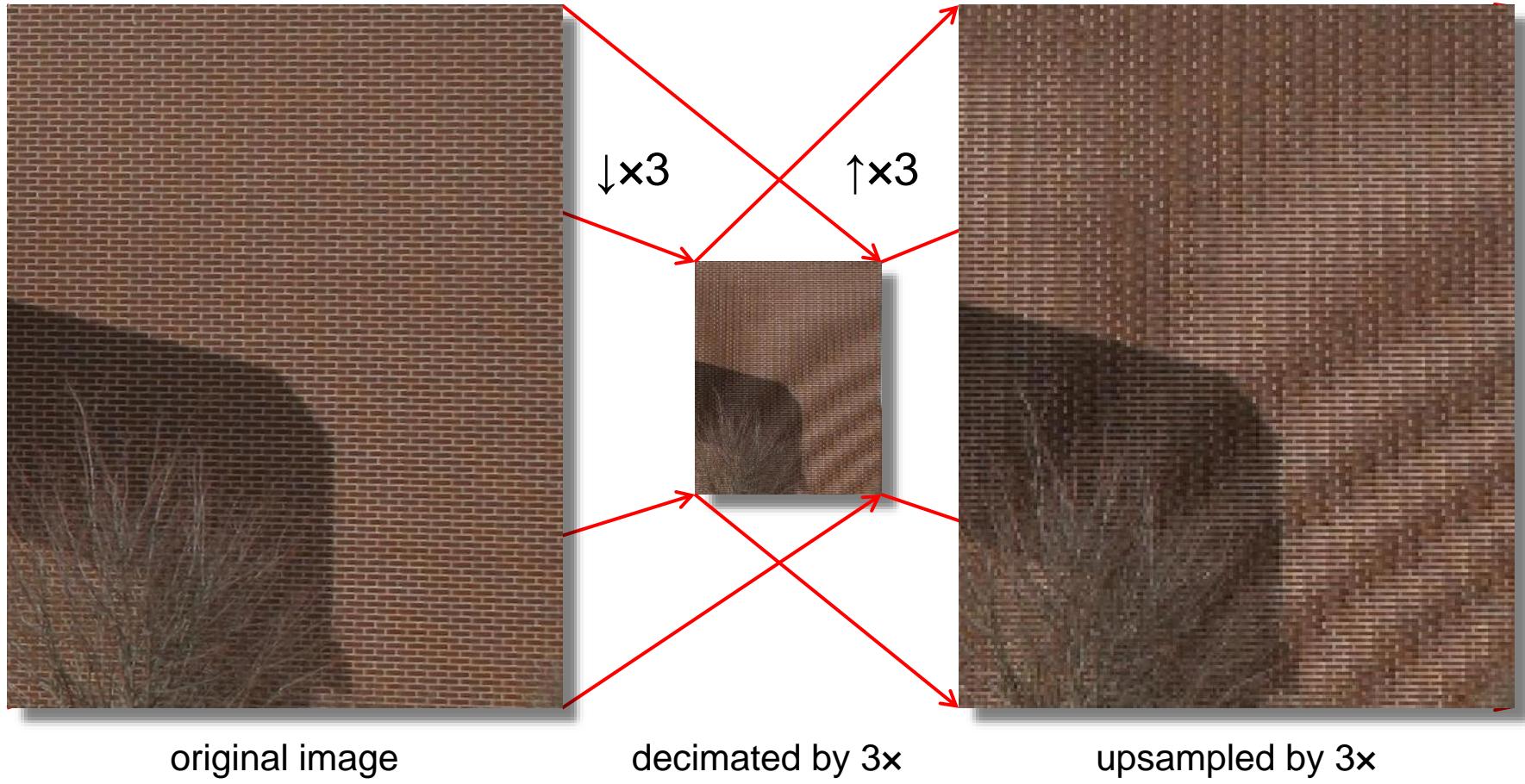


E.g.: every 2nd pixel in every 2nd row



Jaggies and Moire

Another fine mess:
Moire distortion.



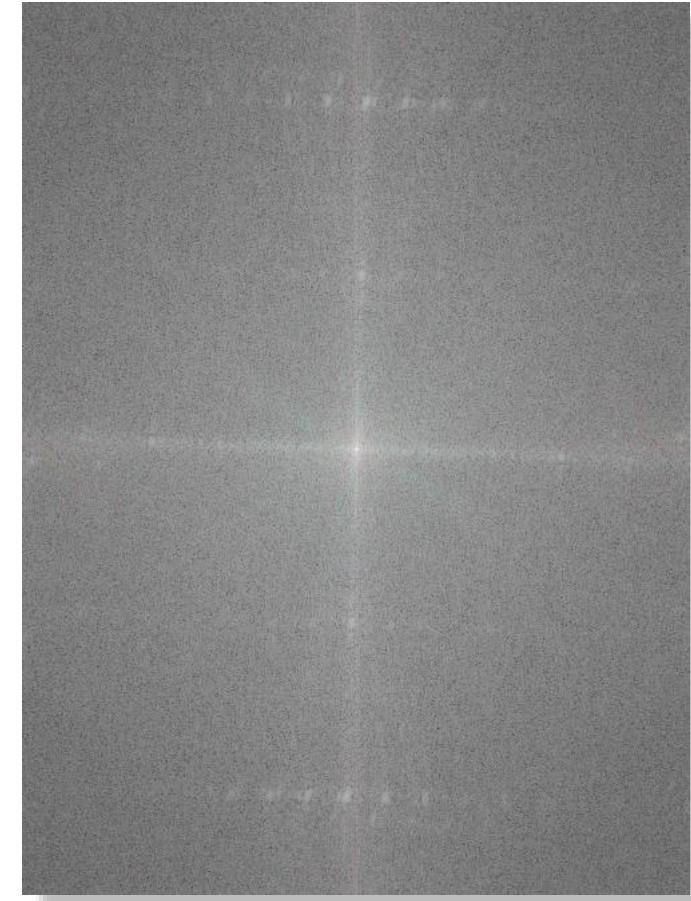


Recall:

Power Spectrum from Discrete Fourier Transform



DFT

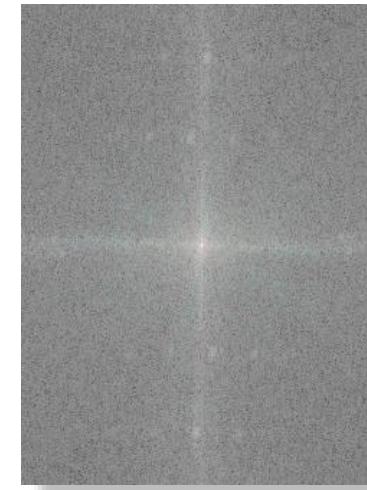




Recall:

Power Spectrum from Discrete Fourier Transform

decimated image



power spectrum

The DFT of an image is the same size as the image.



Recall:

The Scaling Property of the FT

If $\mathcal{F}\{\mathbf{I}\}(v, u) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mathbf{I}(r, c) e^{-i2\pi(uc+vr)} dc dr,$

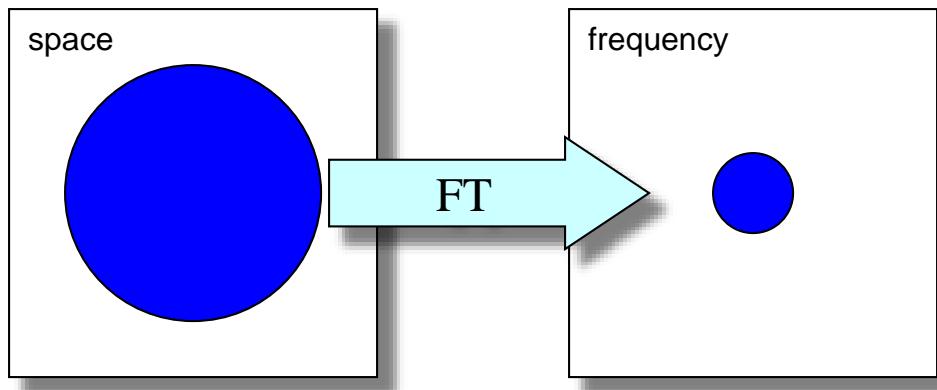
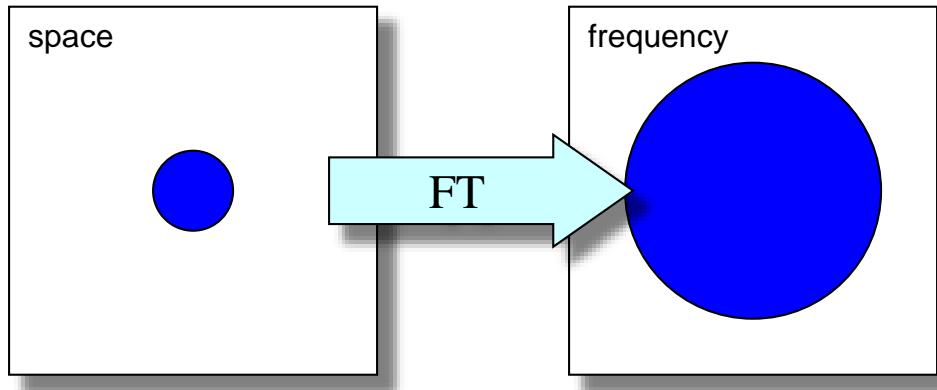
then $\mathbf{I}\left(\frac{r}{a}, \frac{c}{b}\right) \Leftrightarrow |ab| \mathcal{F}\{\mathbf{I}\}(av, bu).$

This implies that if an image is reduced in size, its features in the spatial domain become smaller and its features in the frequency domain become larger.



Recall:

The Uncertainty Relation



If $\Delta x \Delta y$ is the extent of the object in space and if $\Delta u \Delta v$ is its extent in frequency then

$$\Delta x \Delta y \cdot \Delta u \Delta v \geq \frac{1}{16\pi^2}$$

A small object in space has a large frequency extent and vice-versa.



Effect of Decimation on the DFT of an Image

1. Decimation of an $R \times C$ image, \mathbf{I} , by a factor of n results in an $\lfloor R/n \rfloor \times \lfloor C/n \rfloor$ image, \mathbf{J} .
2. The DFT of image \mathbf{J} is the same size as \mathbf{J} .
3. The uncertainty relation implies that the FT of \mathbf{J} should be $(R-1) \times (C-1) < n \lfloor R/n \rfloor \times n \lfloor C/n \rfloor \leq R \times C$.

Q: How can these 3 facts be true simultaneously?

A: The FT of \mathbf{J} *folds over* or *aliases* itself on the DFT of \mathbf{J} because the DFT is defined on a torus.



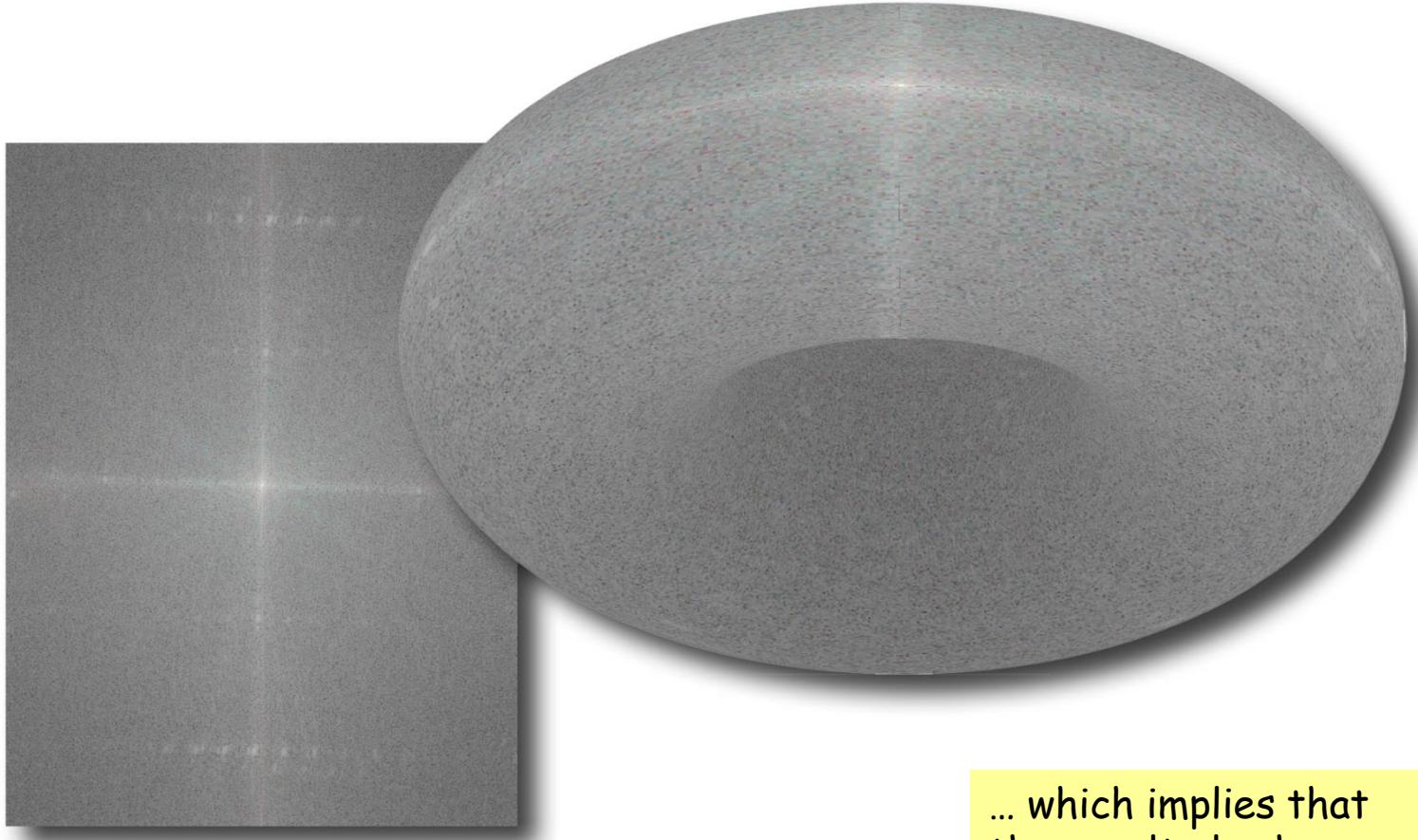
Discrete FT is on a Torus



Recall: the DFT of a digital image assumes the image has a toroidal topology...



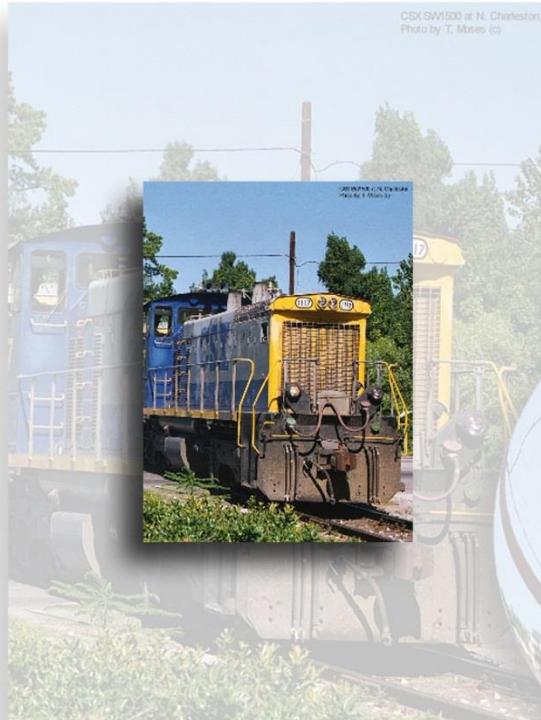
Discrete Fourier Transform is on a Torus



... which implies that
the result also has a
toroidal topology.



Discrete FT is on a Torus

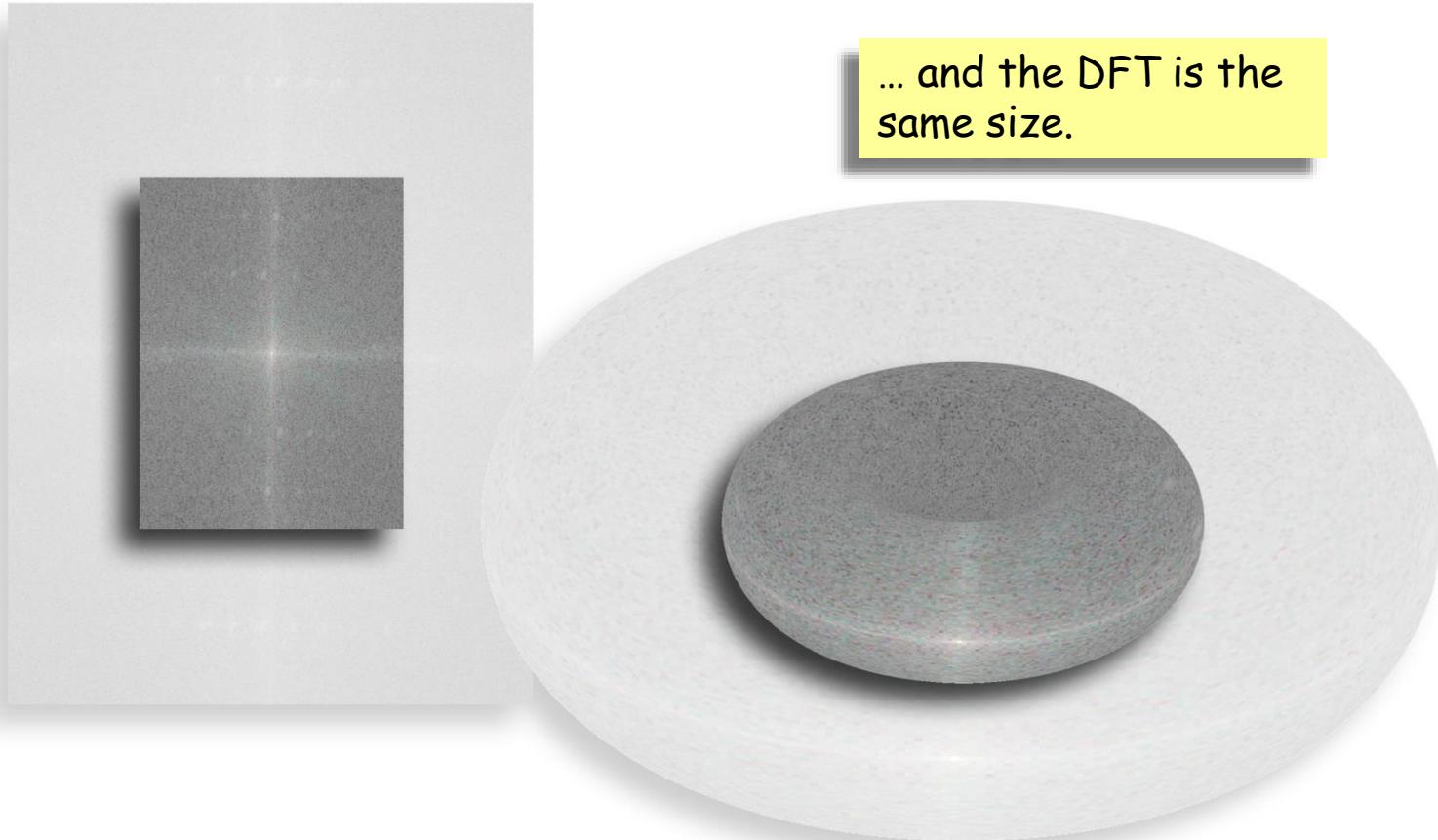


If the image is made smaller the torus is likewise ...





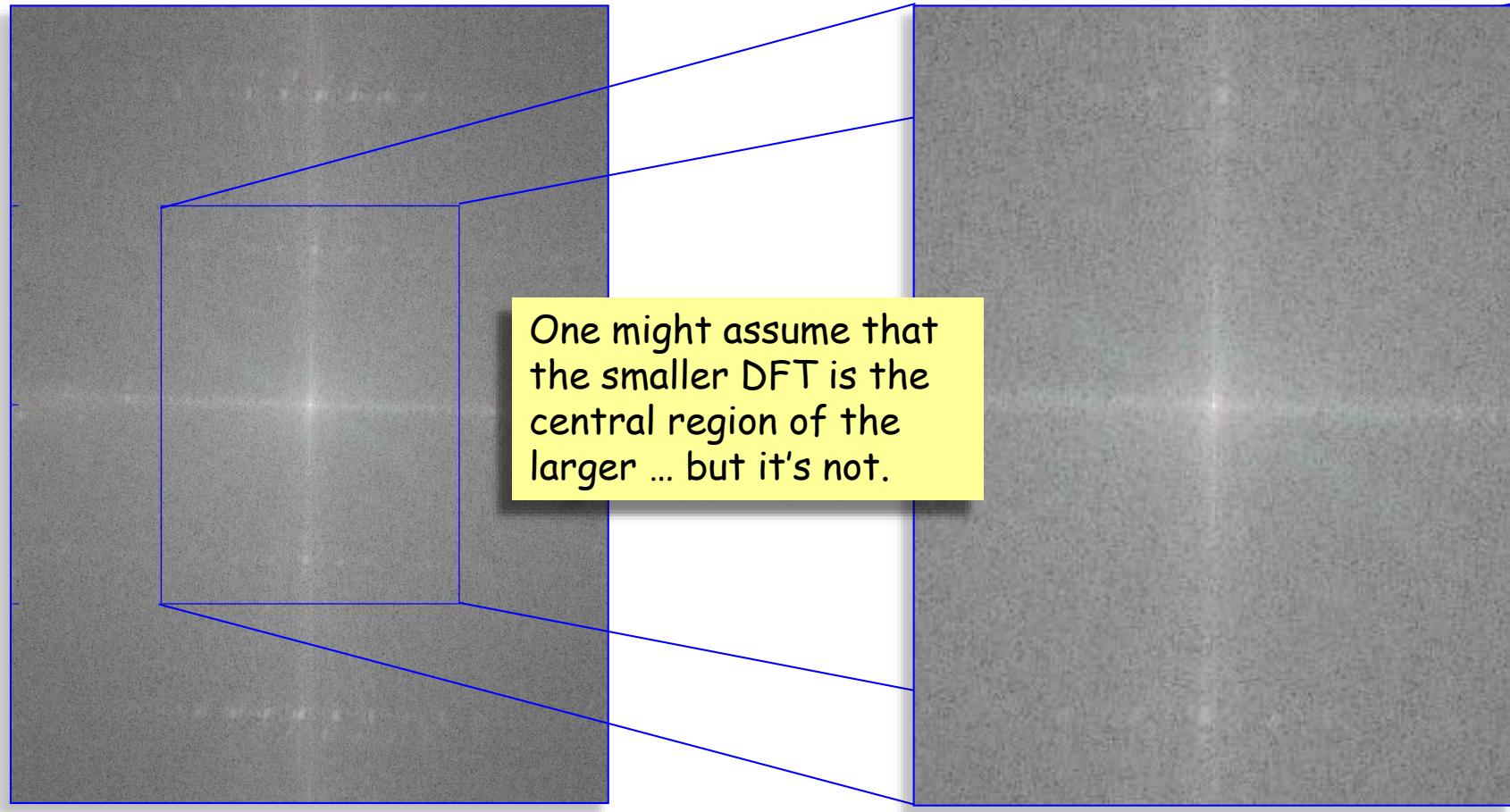
Discrete FT is on a Torus





Ideal PS of $2 \times$ Decimated Image

zoomed $\times 2$

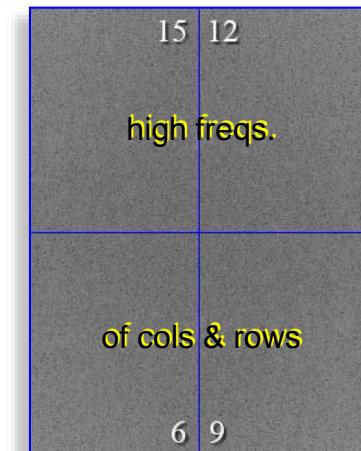
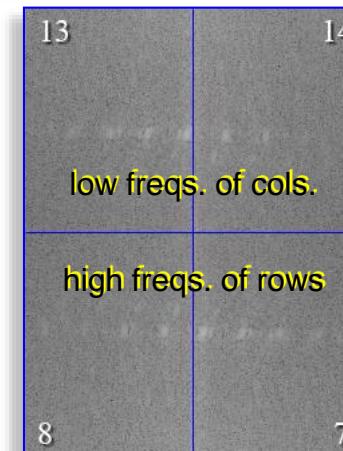
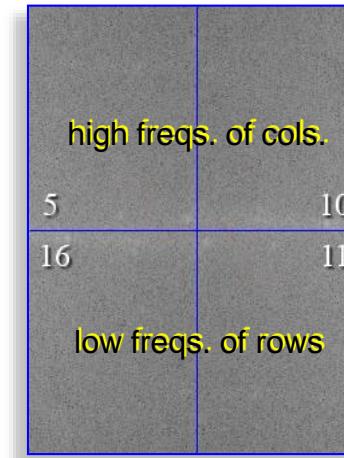
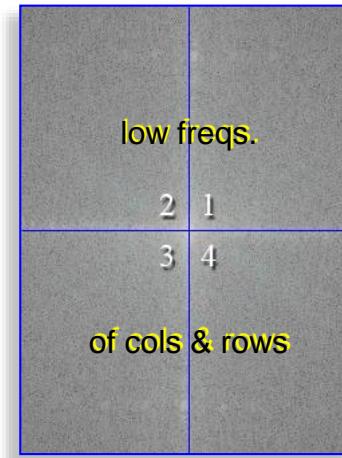




All of the larger one is present
in the smaller one.

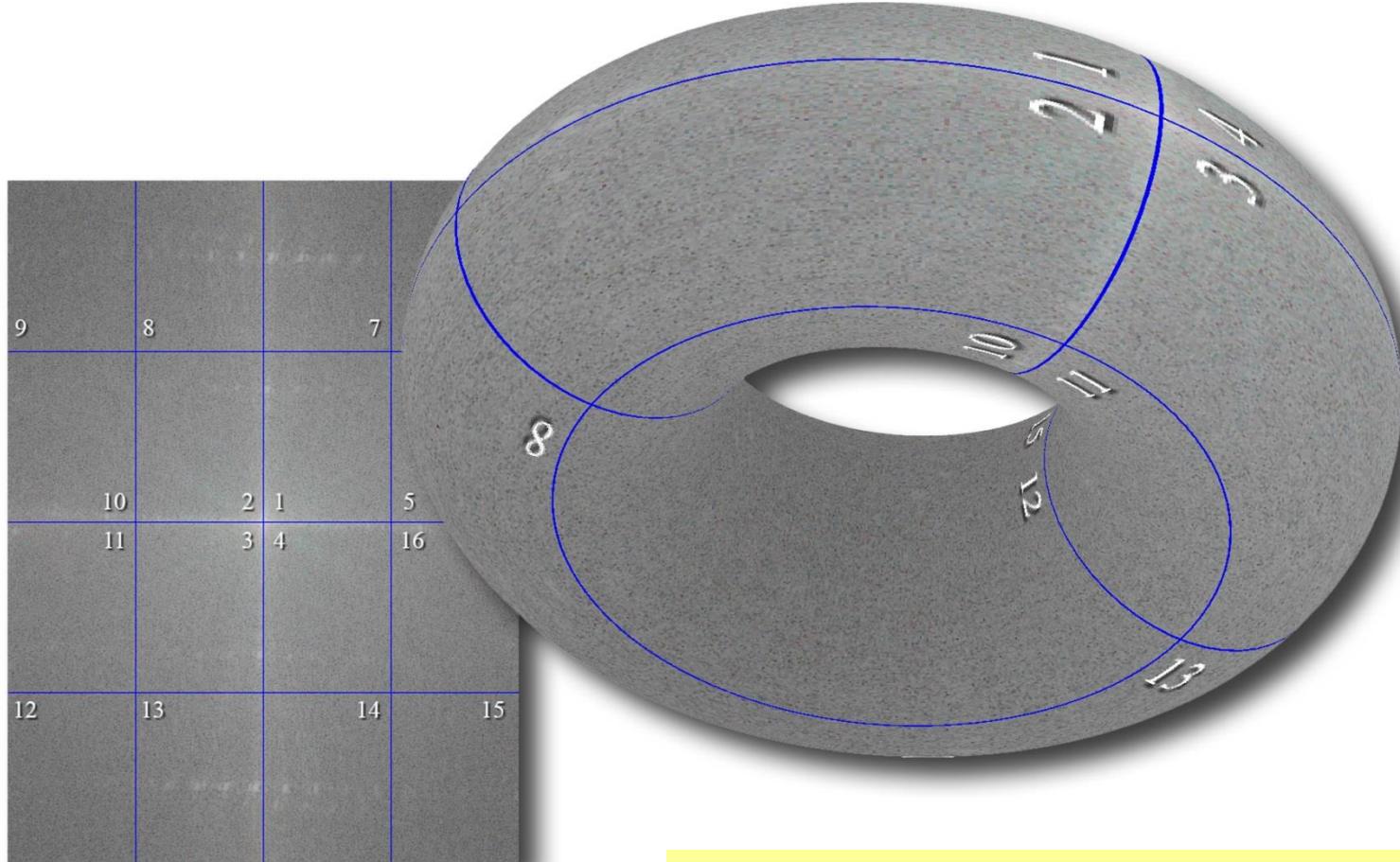
To Make Actual PS of $2 \times$ Decimated Image:

| | | | |
|----|----|----|----|
| 9 | 8 | 7 | 6 |
| 10 | 2 | 1 | 5 |
| 11 | 3 | 4 | 16 |
| 12 | 13 | 14 | 15 |





Actual PS of $2 \times$ Dec:

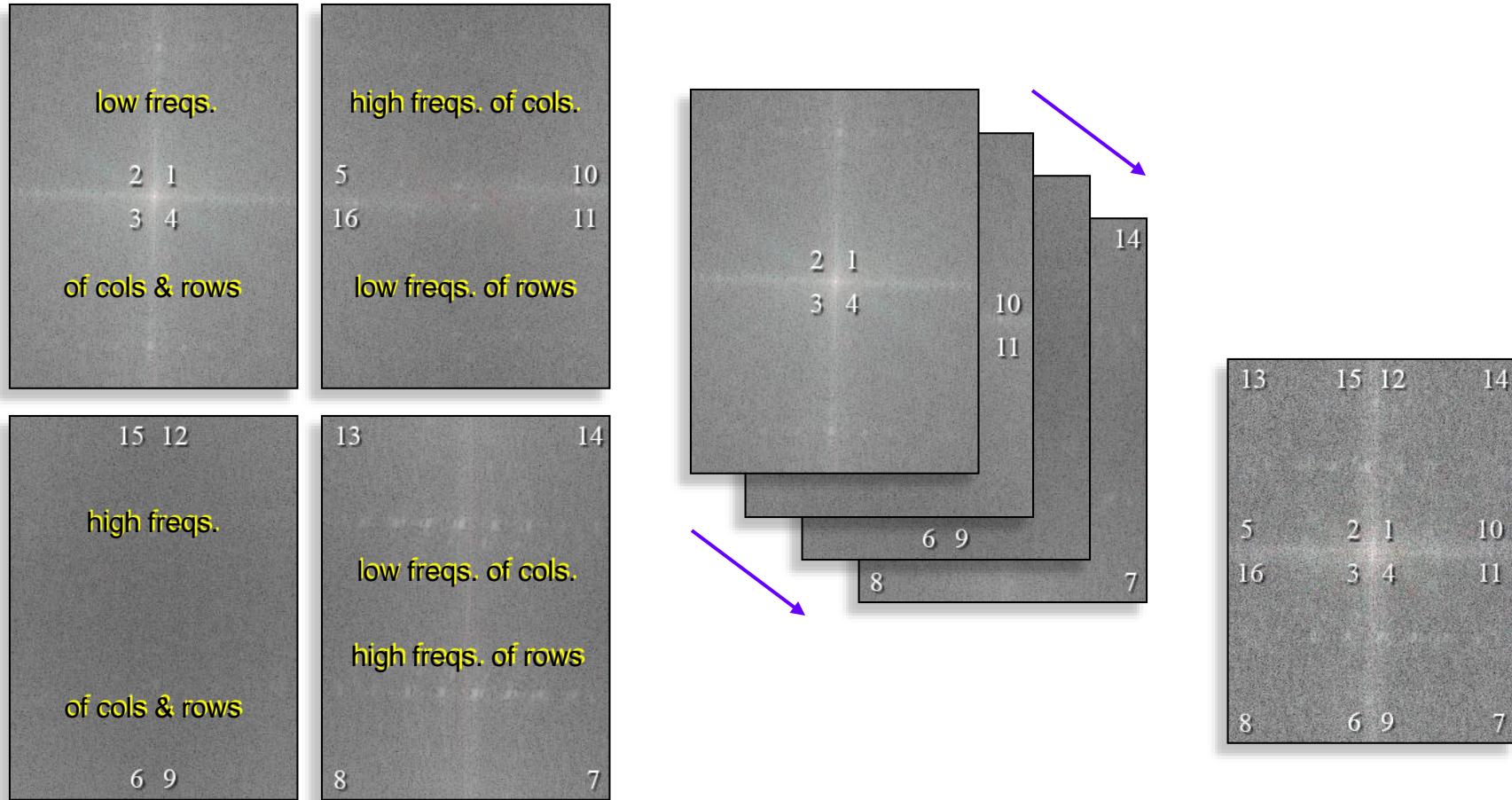


... four parts that are contiguous
on the full-sized DFT torus.



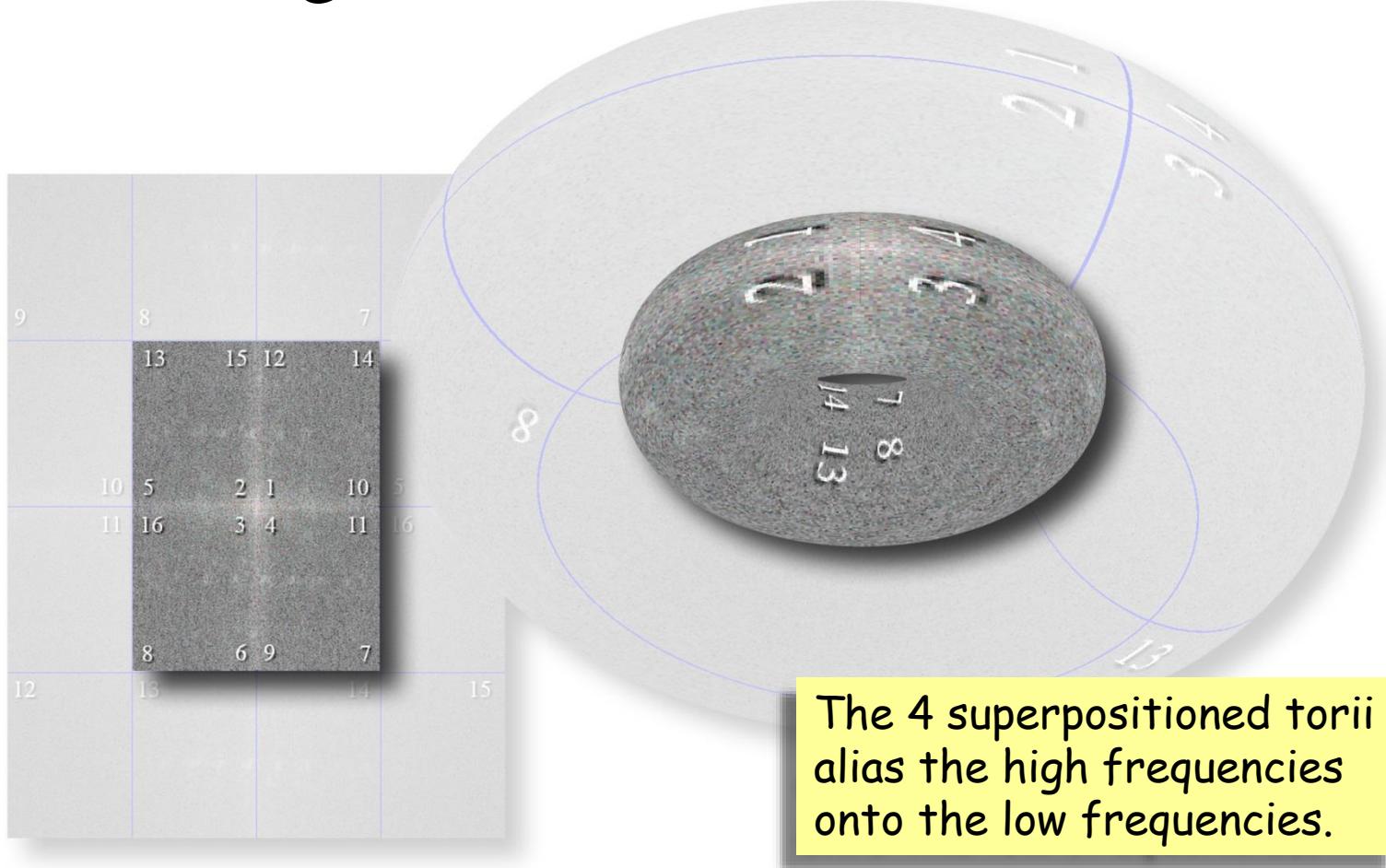
Each of the 4 PS regions forms a torus. The 4 torii are superpositioned onto 1.

Actual PS of $2 \times$ Decimated Image



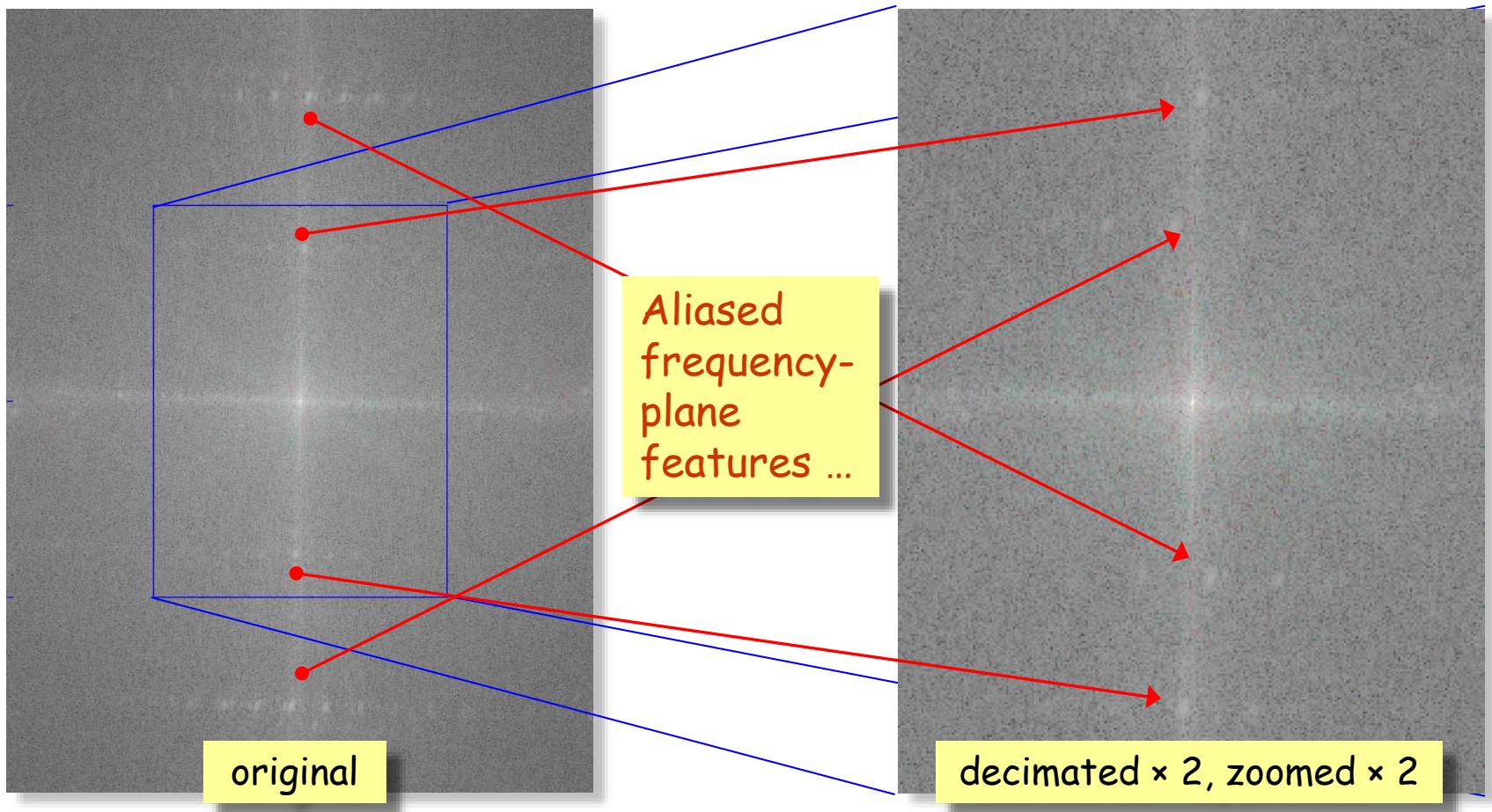


DFT Aliasing on the Torus





PS of Original and PS of $2 \times$ Decimated Image





... lead to jaggies in
the decimated image.

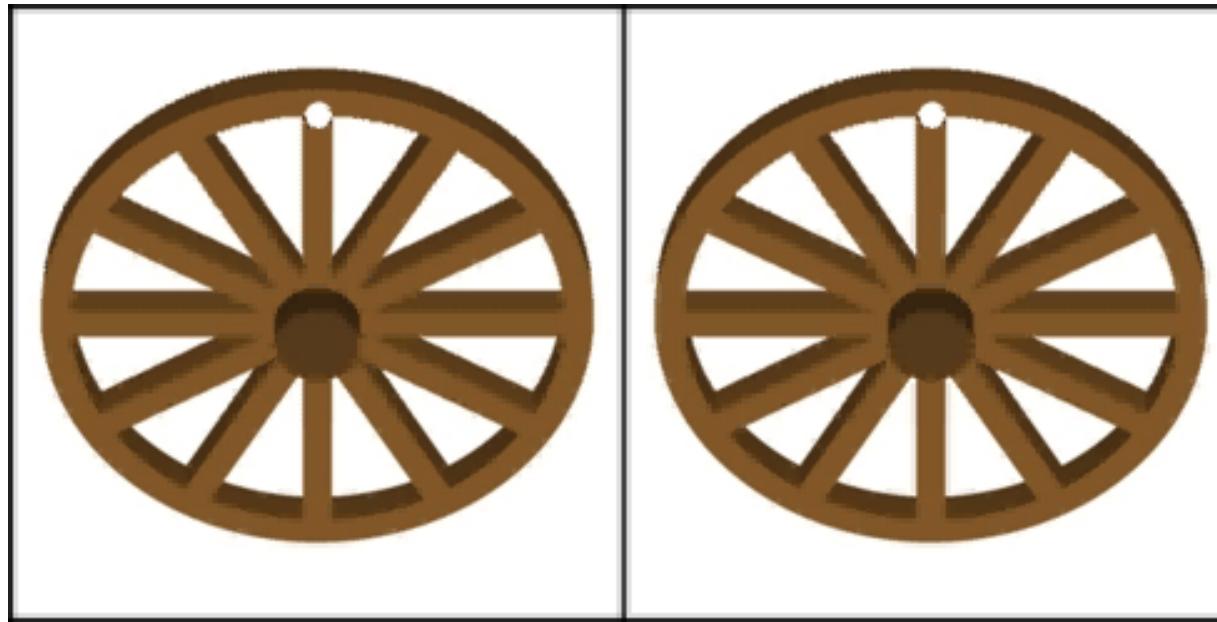
Original Image and $2\times$ Decimated Image





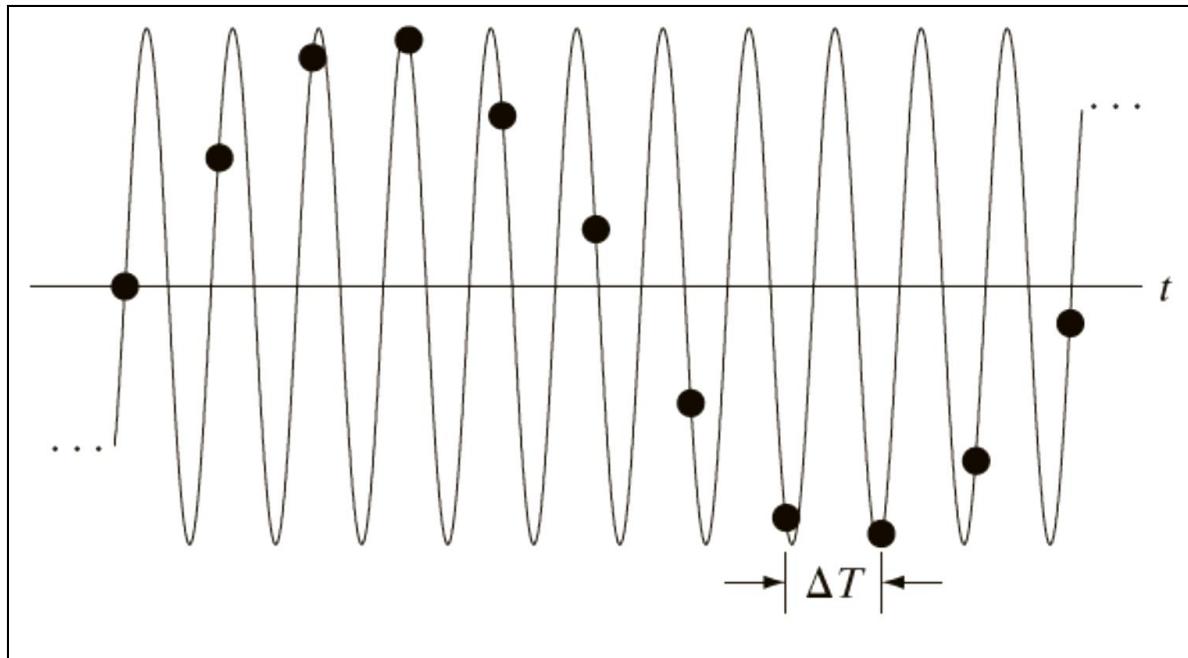
First Alternative Explanation of Aliasing

The aliasing phenomenon can be described in terms of the so-called “wagon wheel” effect. The name comes from the appearance in a 24 frame per second movie of a wagon wheel rotating slightly faster than 12 frames per second. It appears to be rotating slowly backward.





Signal Aliasing



An under-sampled sinusoid is a sinusoid of a different frequency.

Figure from George Bebis, Univ. of Nevada, Reno, CS474, Image Processing and Interpretation, Lecture 4.3.



Nyquist-Shannon & Whittaker-Shannon Sampling

If a function [signal or image] $x(t)$ contains no frequencies higher than B hertz, it is completely determined by giving its ordinates at a series of points spaced $1/(2B)$ seconds apart. In other words, a band-limited function can be perfectly reconstructed from a countable sequence of samples if the band limit, B , is no greater than half the sampling rate (samples per second).¹

Given a sequence of real numbers, $x[n]$, the continuous function:

$$x(t) = \sum_{n=-\infty}^{\infty} x[n] \cdot \text{sinc}\left(\frac{t - nT}{T}\right)$$

has a Fourier transform, $X(f)$, whose non-zero values are confined to the region: $|f| \leq 1/2T$. When parameter T has units of seconds, the band limit, $1/2T$, has units of cycles/sec (hertz). When the $x[n]$ sequence represents time samples, at interval T , of a continuous function, the quantity $f_s = 1/T$ is known as the sample rate, and $f_s/2$ is the corresponding Nyquist frequency. When the sampled function has a bandlimit, B , less than the Nyquist frequency, $x(t)$ is a perfect reconstruction of the original function.²

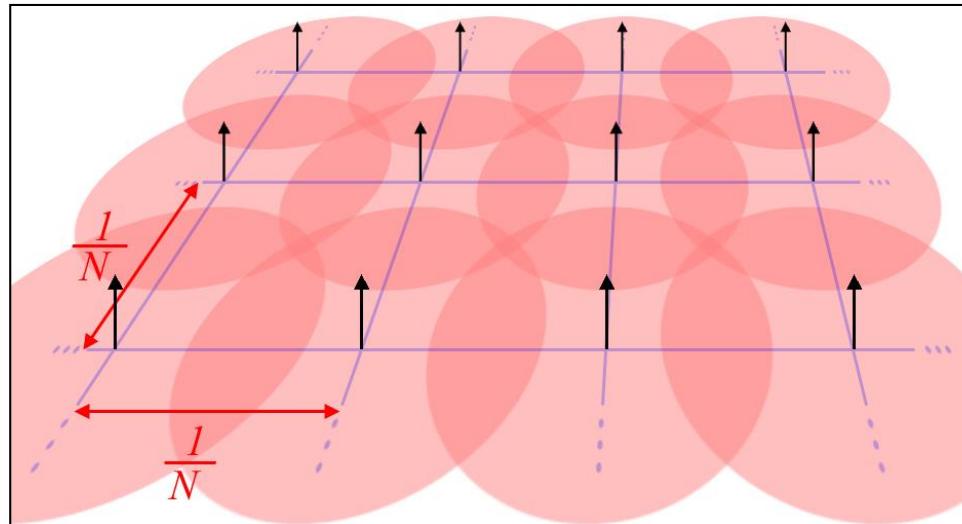
¹http://en.wikipedia.org/wiki/Nyquist–Shannon_sampling_theorem

²http://en.wikipedia.org/wiki/Whittaker–Shannon_interpolation_formula



Second Alternative Explanation of Aliasing

The aliasing phenomenon can also be described in terms of the convolution property of the Fourier Transform. In fact, this is the way it is usually explained.

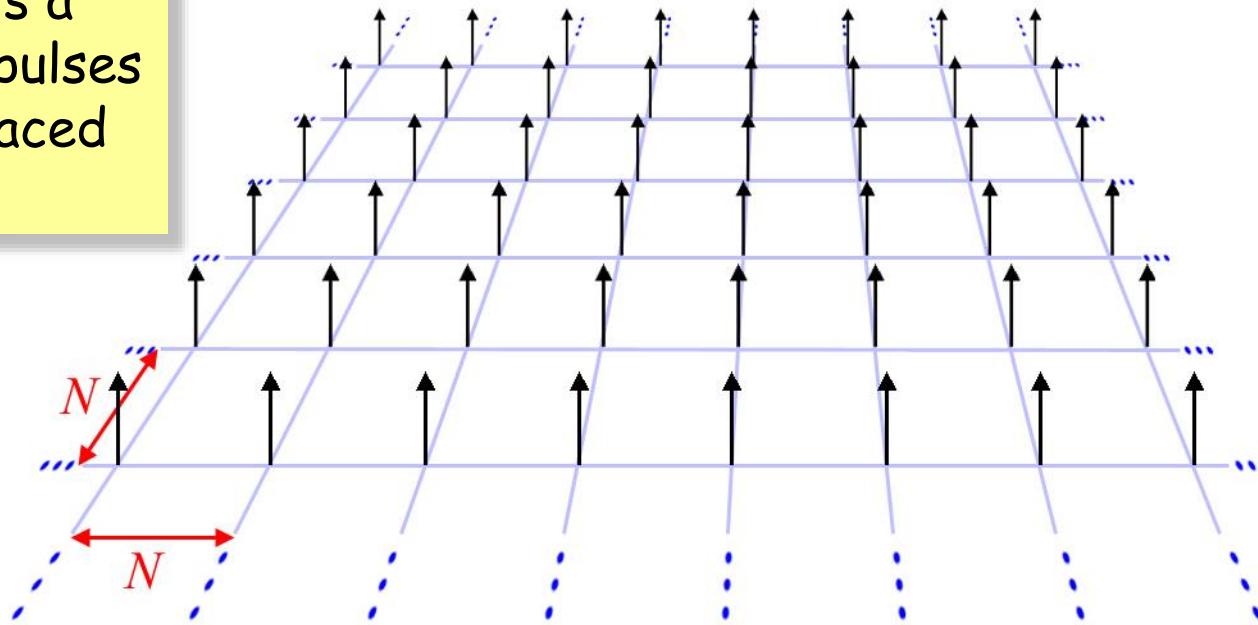




The Sampling Function

The sampling function is a set of impulses evenly spaced on a grid.

$$\text{samp}_N(r, c) = \sum_{j=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} \delta(r - jN) \delta(c - kN)$$



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The Sampling of an Image

An image is sampled by multiplying it by the sampling function

$$I(r, c) \cdot \text{samp}_N(r, c)$$

$$\text{samp}_N(r, c) = \sum_{j=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} \delta(r - jN) \delta(c - kN)$$

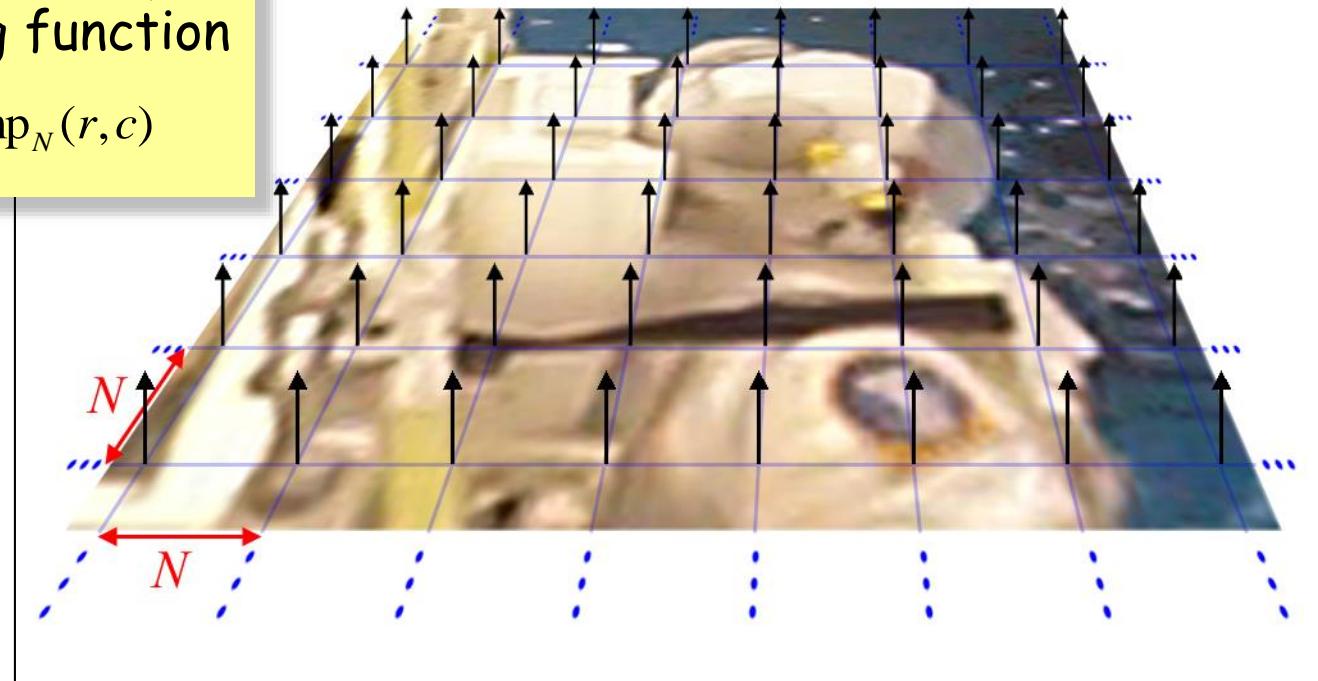
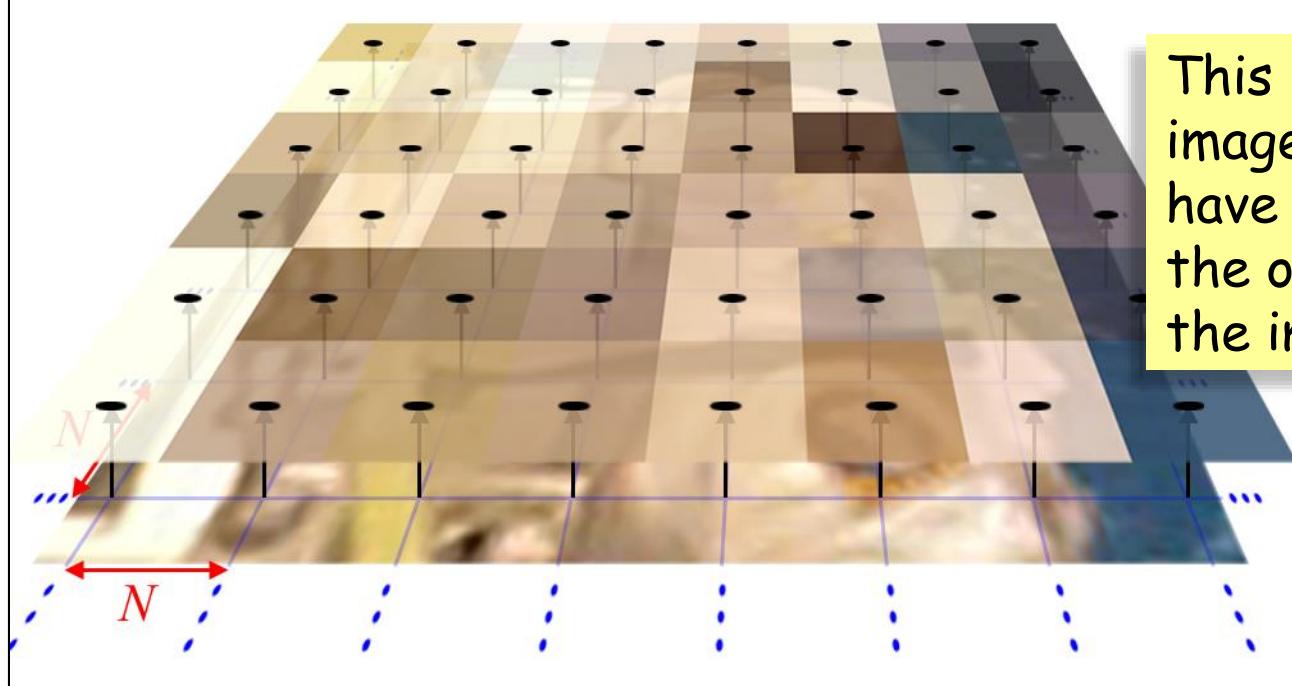




Image Sampling in the Spatial Domain

$$\text{samp}_N \{ \mathbf{I} \}(r, c) = \mathbf{I}(r, c) \cdot \sum_{j=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} \delta(r - jN) \cdot \delta(c - kN)$$



This results in an image whose pixels have the values of the original image at the impulse locations.



Convolution Property of the Fourier Transform

Let functions $f(r, c)$ and $g(r, c)$ have

Fourier Transforms $F(u, v)$ and $G(u, v)$.

Then,

$$\mathfrak{F}\{f * g\} = F \cdot G.$$

Moreover,

$$\mathfrak{F}\{f \cdot g\} = F * G.$$

* represents convolution

· represents pointwise multiplication

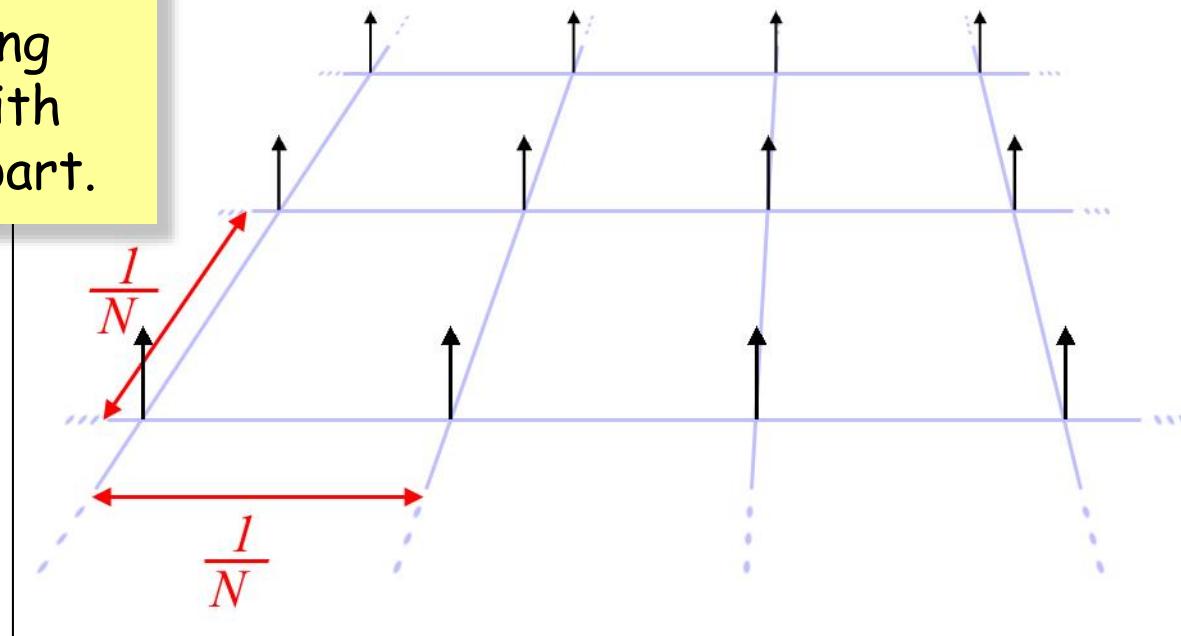
The Fourier Transform of a product equals the convolution of the Fourier Transforms.
Similarly, the Fourier Transform of a convolution is the product of the Fourier Transforms



The Fourier Transform of the Sampling Function

The Fourier Transform of the sampling function is another sampling function but with impulses $1/N$ apart.

$$\text{samp}_{1/N}(u,v) = \sum_{j=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} \delta(u - \frac{j}{N}) \delta(v - \frac{k}{N})$$



Cf. slide 23



Convolution by an Impulse

An *impulse* is a digital image, that has a single pixel with value 1; all others have value zero. An impulse at location (ρ, χ) is represented by:

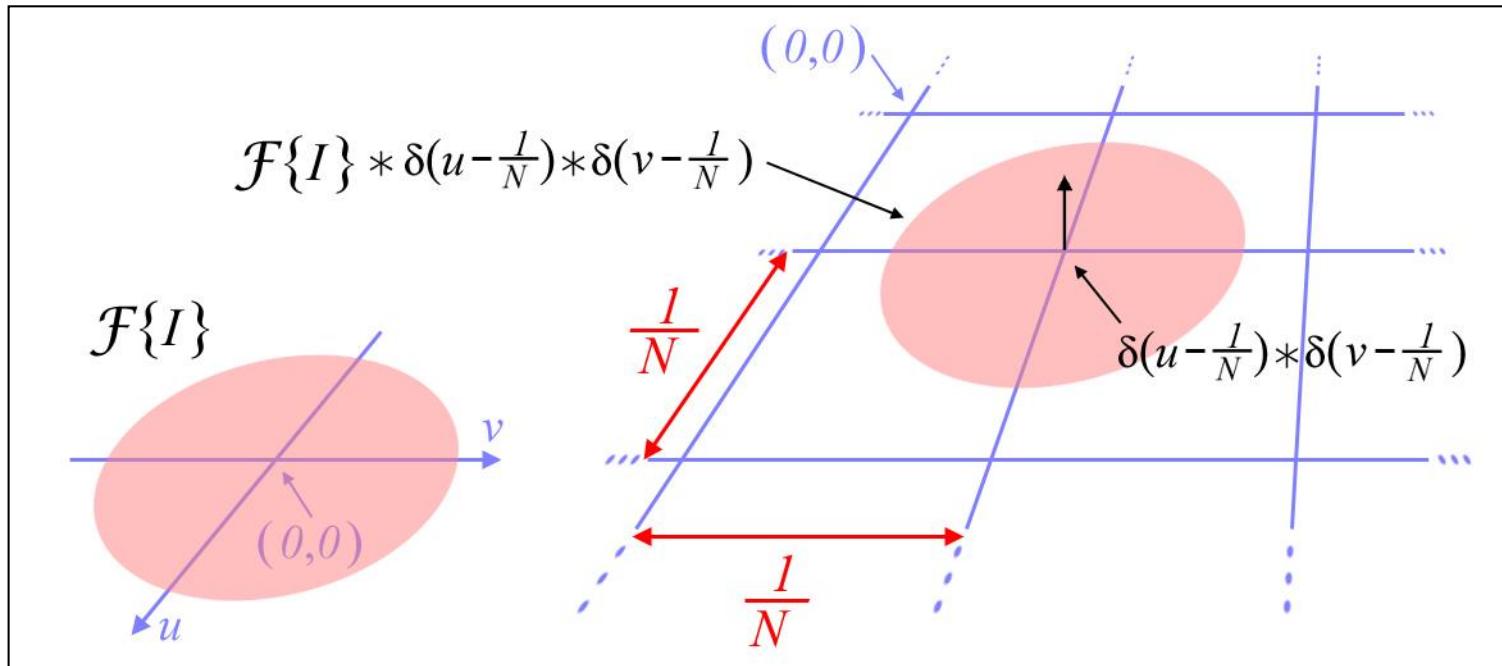
$$\delta(r - \rho, c - \chi) = \begin{cases} 1, & \text{if } r = \rho \text{ and } c = \chi \\ 0, & \text{otherwise} \end{cases}$$

If an image is convolved with an impulse at location (ρ, χ) , the image is shifted in location down by ρ pixels and to the right by χ pixels.

$$[\mathbf{I} * \delta(r - \rho, c - \chi)](r, c) = \mathbf{I}(r - \rho, c - \chi).$$



Convolution by an Impulse



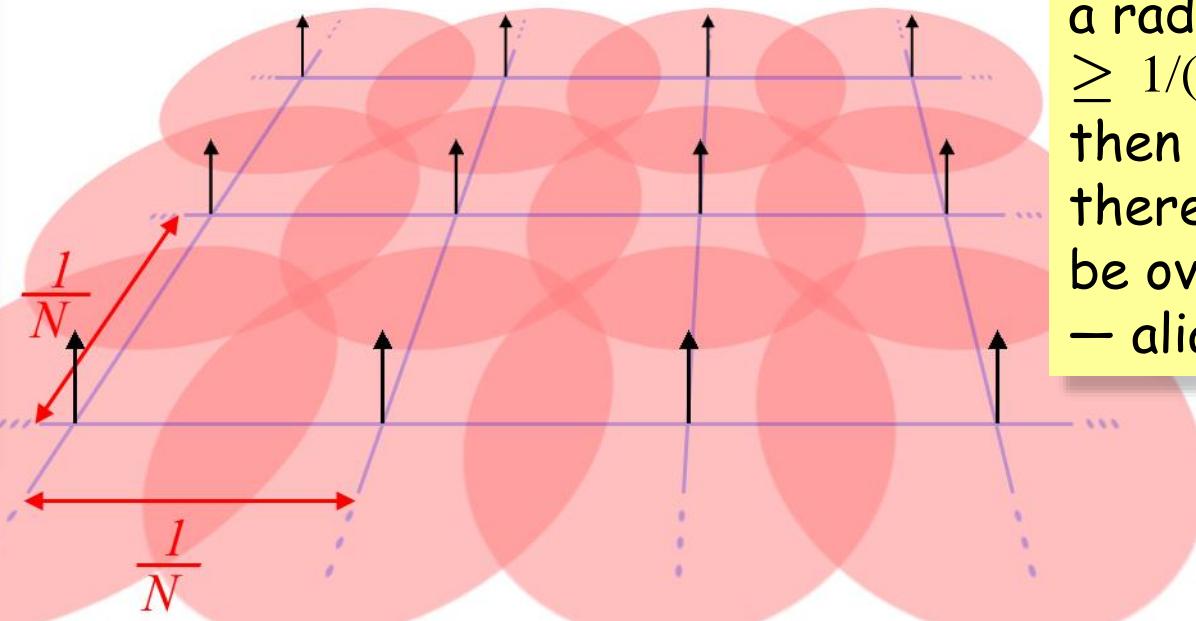
The convolution of any function with a delta function translates the function to the location of the impulse.



The Fourier Transform of a Sampled Image

The sampling of image \mathbf{I} causes its Fourier Transform $\mathcal{F}\{\mathbf{I}\}$ to be repeated at intervals of $1/N$.

$$\sum_{j=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} \mathcal{F}\{\mathbf{I}\}(u, v) * \delta\left(u - \frac{j}{N}\right) * \delta\left(v - \frac{k}{N}\right)$$



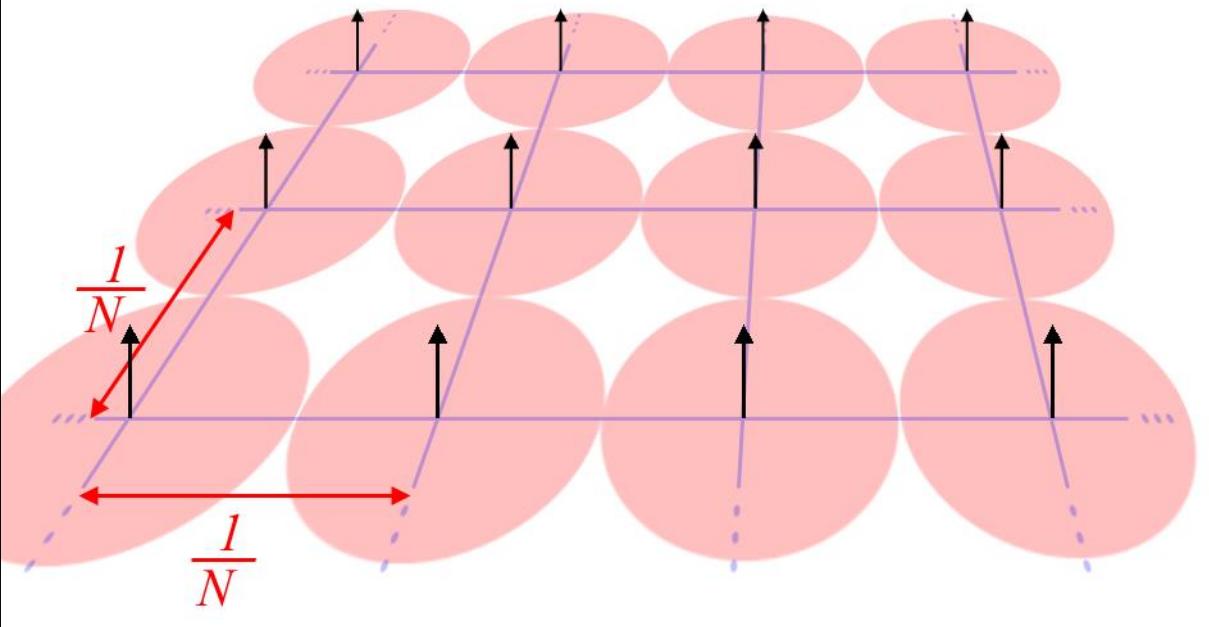
If the support of $\mathcal{F}\{\mathbf{I}\}$ has a radius $\geq 1/(2N)$ then there will be overlap — aliasing



The FT of a Sampled Image with No Aliasing

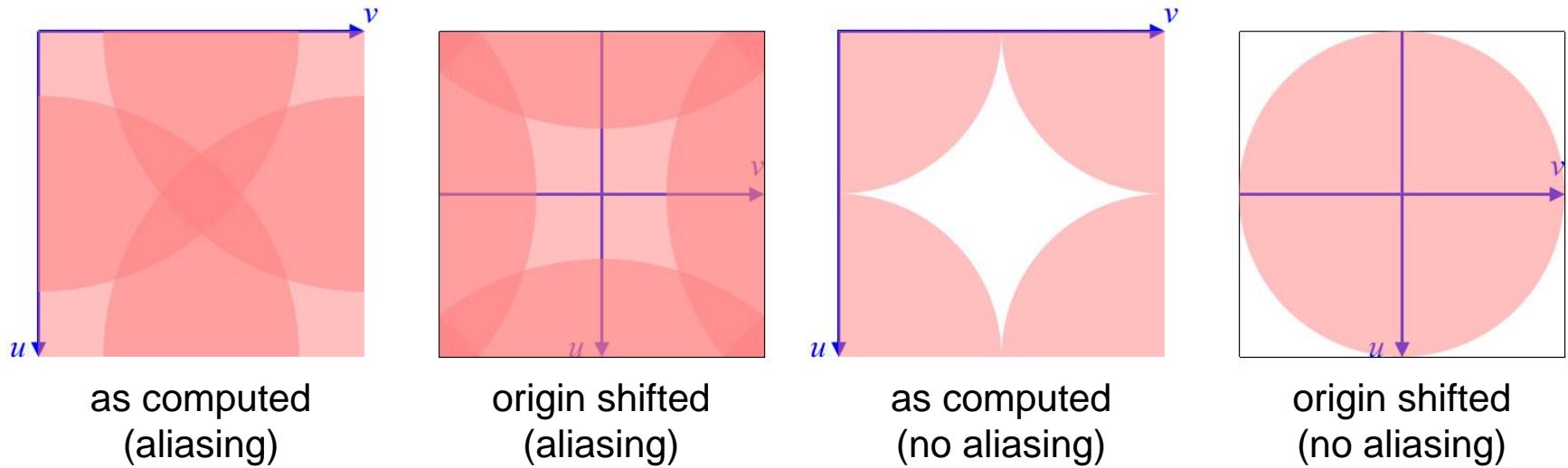
For aliasing not to occur the frequency support of \mathbf{I} must have a radius of $< 1/(2N)$

$$\sum_{j=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} \mathfrak{F}\{\mathbf{I}\}(u, v) * \delta\left(u - \frac{j}{N}\right) * \delta\left(v - \frac{k}{N}\right)$$





Part of the Transform Computed by the FFT



Although the FT of the sampled image continues on indefinitely in theory, only one copy of the complete pattern is required for processing. The Fast Fourier Transform (FFT) algorithm computes the transform with the origin in the upper left. Usually the transform is displayed with the origin in the center.

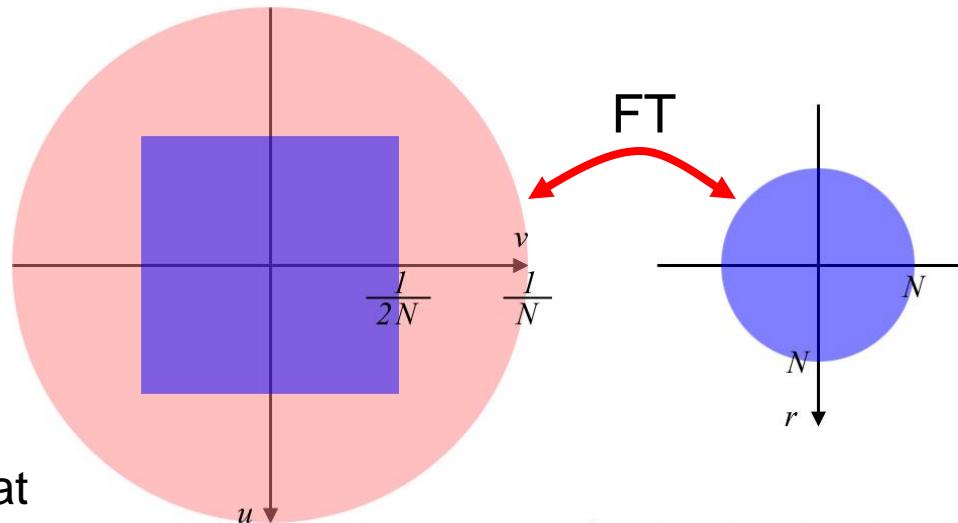


Filtering Before Downsampling to Prevent Aliasing

To sample $\mathbf{I}(r,c)$ every N pixels multiply $\mathcal{F}(u,v) = \mathcal{F}\{\mathbf{I}\}$ by a window $\mathbf{H}(u,v)$ such that

$$\mathbf{H}(u,v) = 0 \text{ for } u > \frac{1}{2N} \text{ or } v > \frac{1}{2N}$$

That is equivalent to convolving $\mathbf{I}(r,c)$ with a spatial filter of width $2N$.



| | | | | |
|----------|----------|----------|----------|----------|
| h_{11} | h_{12} | h_{13} | h_{14} | h_{15} |
| h_{21} | h_{22} | h_{23} | h_{24} | h_{25} |
| h_{31} | h_{32} | h_{33} | h_{34} | h_{35} |
| h_{41} | h_{42} | h_{43} | h_{44} | h_{45} |
| h_{51} | h_{52} | h_{53} | h_{54} | h_{55} |

Thus, to down-sample $\mathbf{I}(r,c)$ by a factor of 2, use a spatial filter of width 4. Since this is midway between 3 and 5, use a 5x5 convolution mask.

If the image is a bit blurry then a 3x3 will usually suffice



Filtering to Prevent Aliasing

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original image



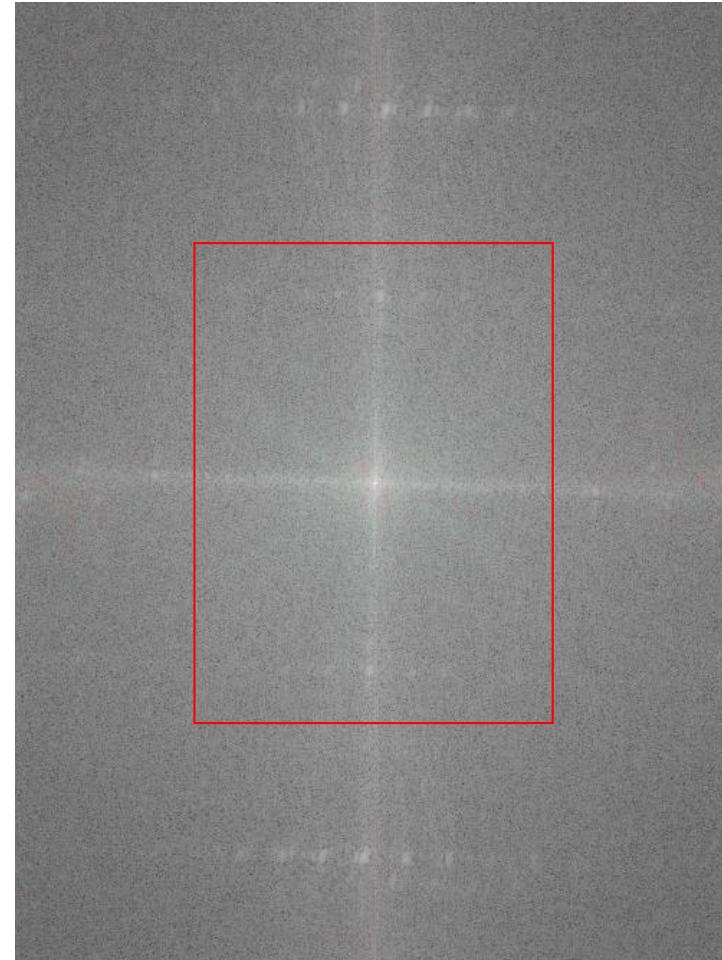
power spectrum



Filtering to Prevent Aliasing



to decimate original by 2 ...



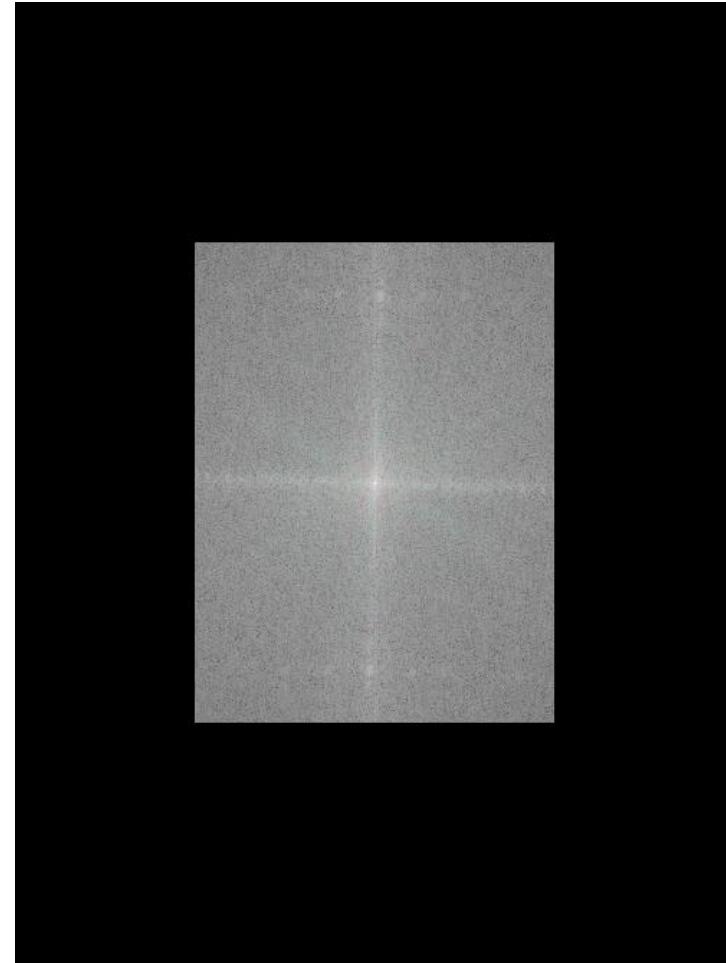
... zero outside red square.



Filtering to Prevent Aliasing



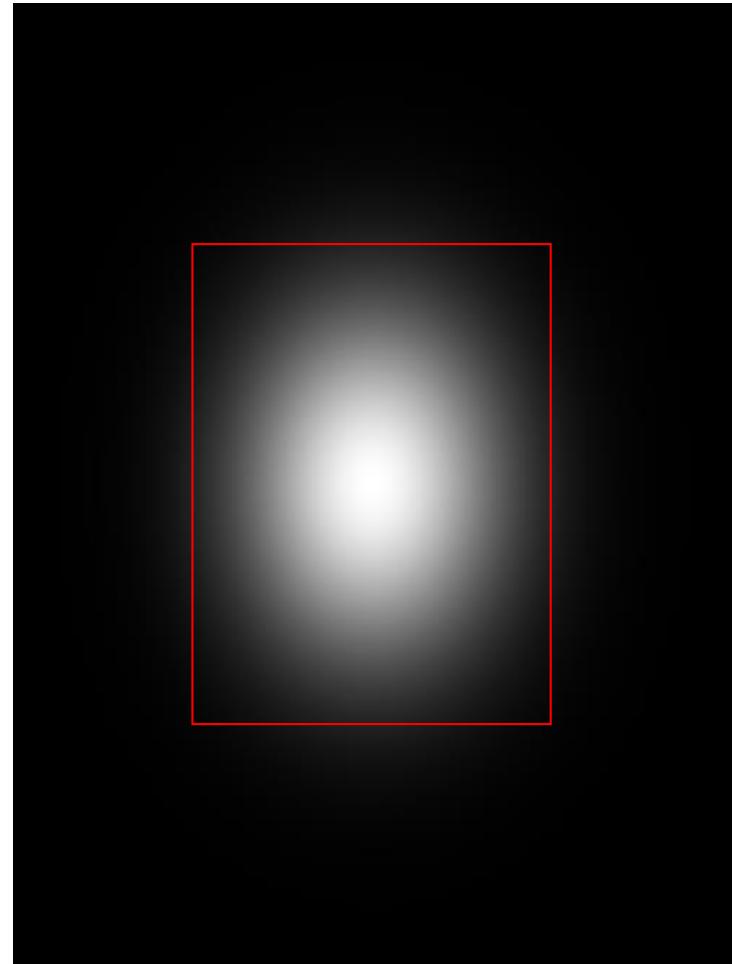
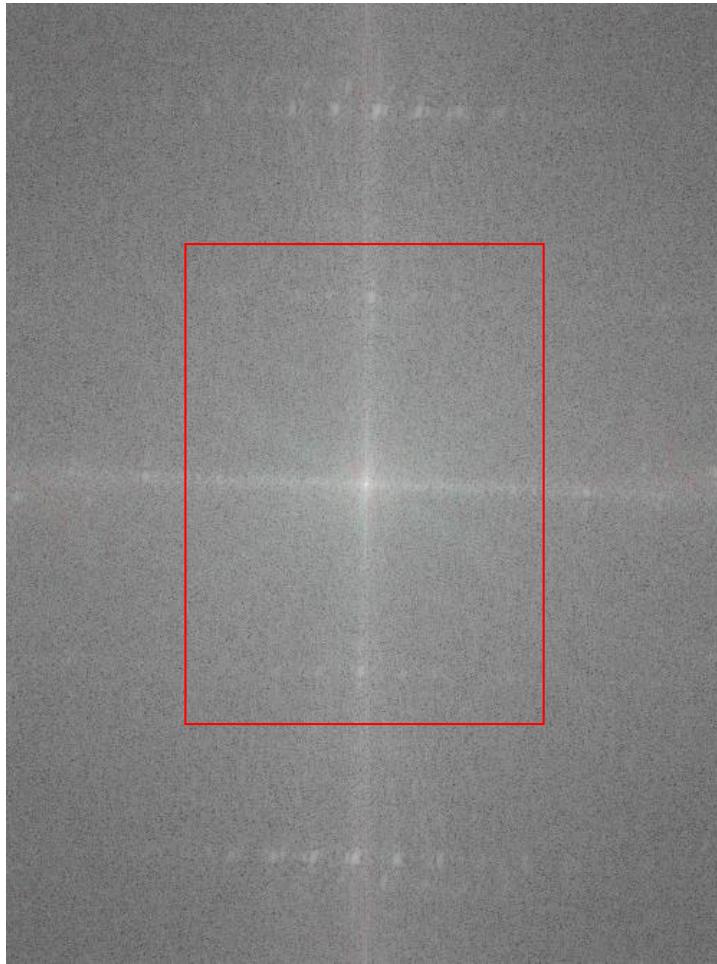
Notice the shift & ringing in the result



ideal filter



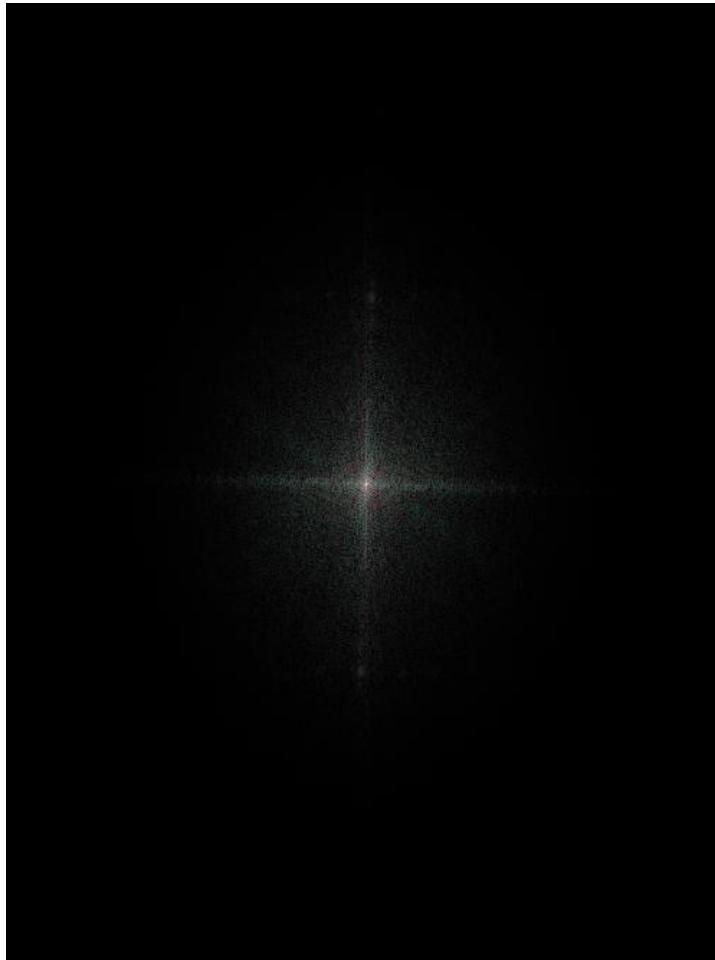
Filtering to Prevent Aliasing



To reduce ringing try multiplying the FT by a Gaussian w/ $(\sigma_v, \sigma_u) = (\frac{1}{4}R, \frac{1}{4}C)$



Filtering to Prevent Aliasing



PS of FT \times Gaussian

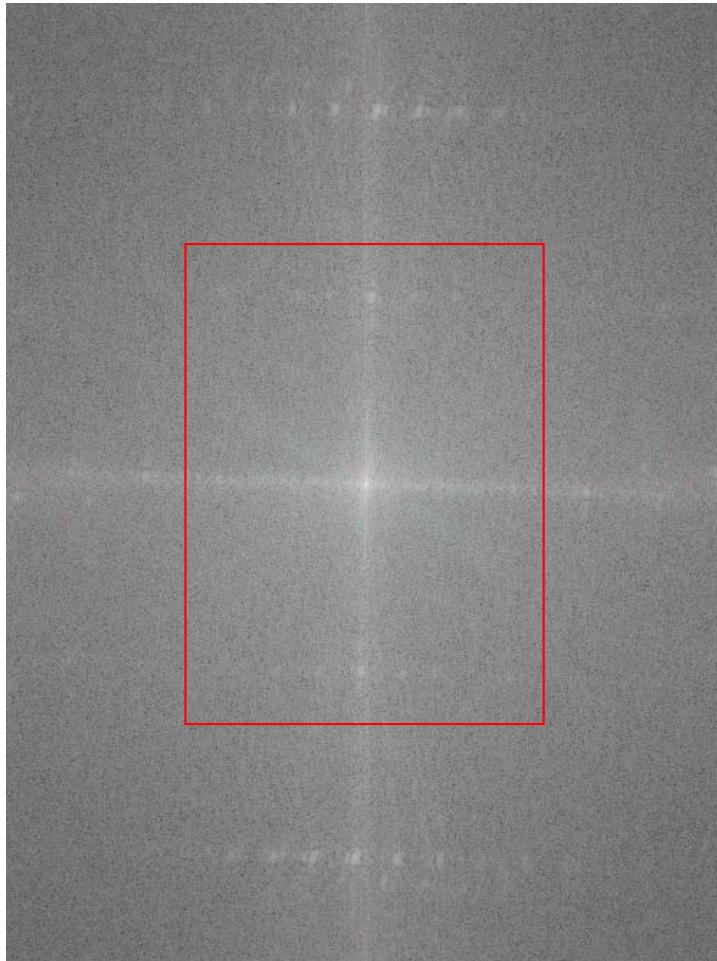


zoomed $\times 2$

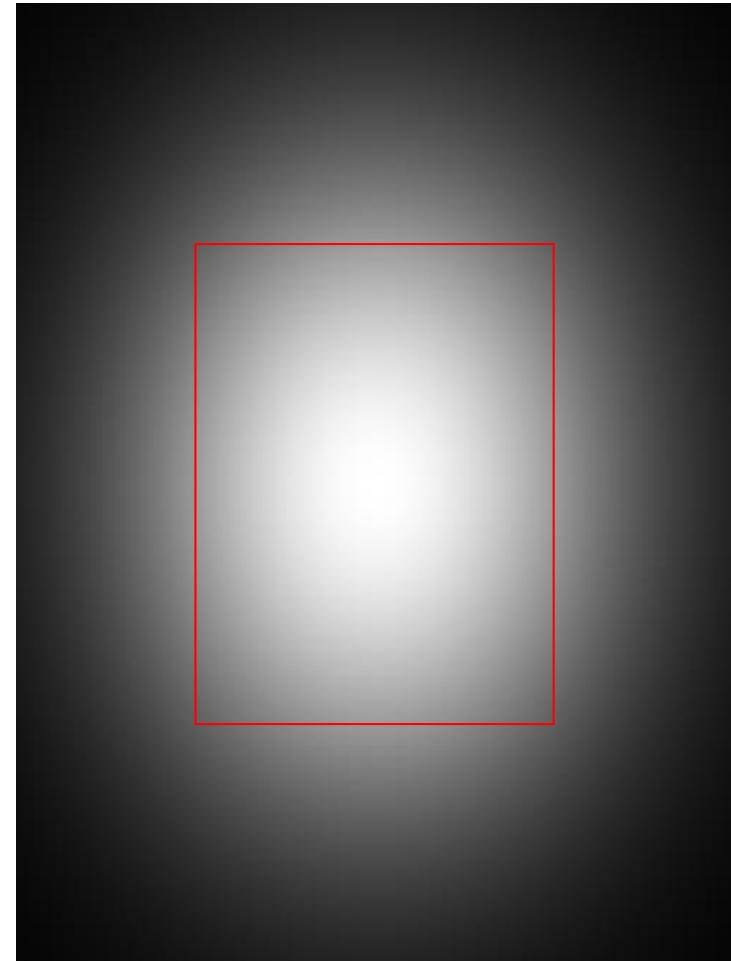
ringing is reduced but result is blurry



Filtering to Prevent Aliasing



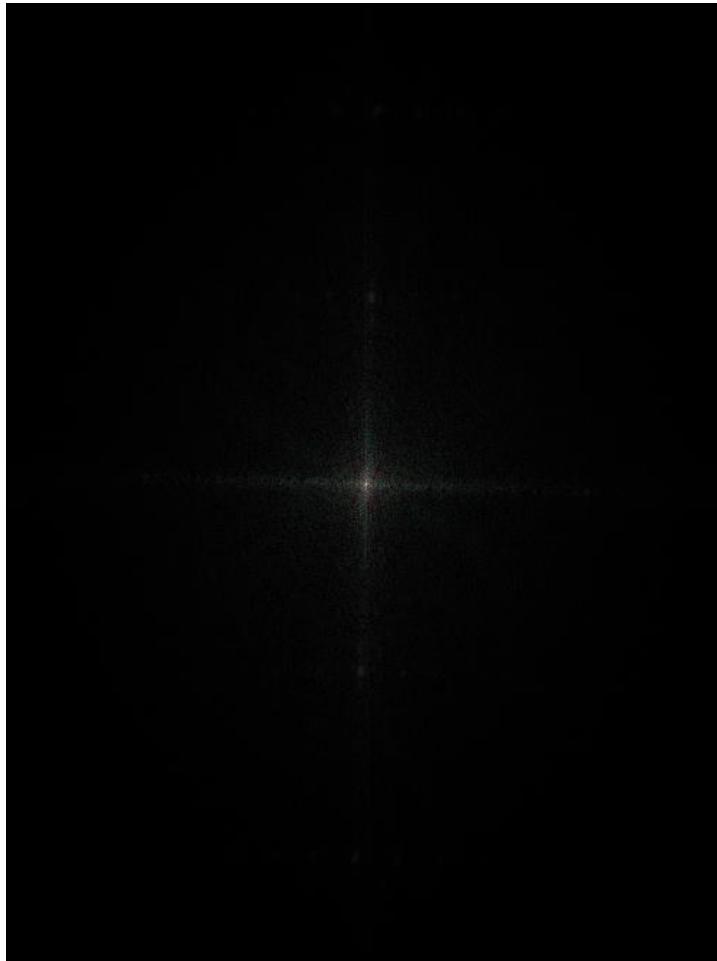
try multiplying the FT



by a Gaussian w/ $(\sigma_v, \sigma_u) = (\frac{1}{2}R, \frac{1}{2}C)$



Filtering to Prevent Aliasing



PS of FT \times Gaussian



some aliasing but result is less blurry



Filtering to Prevent Aliasing



ideal filtered



original



Filtering to Prevent Aliasing

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original



zoomed $\times 2$

Gaussian w/ $(\sigma_v, \sigma_u) = (\frac{1}{4}R, \frac{1}{4}C)$



Filtering to Prevent Aliasing



Gaussian w/ $(\sigma_v, \sigma_u) = (\frac{1}{2}R, \frac{1}{2}C)$





Filtering to Prevent Aliasing

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original



zoomed $\times 2$
decimated by factor of 2



Reconstruction of a Downsampled Image

A band-limited image that was downsampled by a factor of two can be reconstructed – upsampled – using the Fourier Transform.

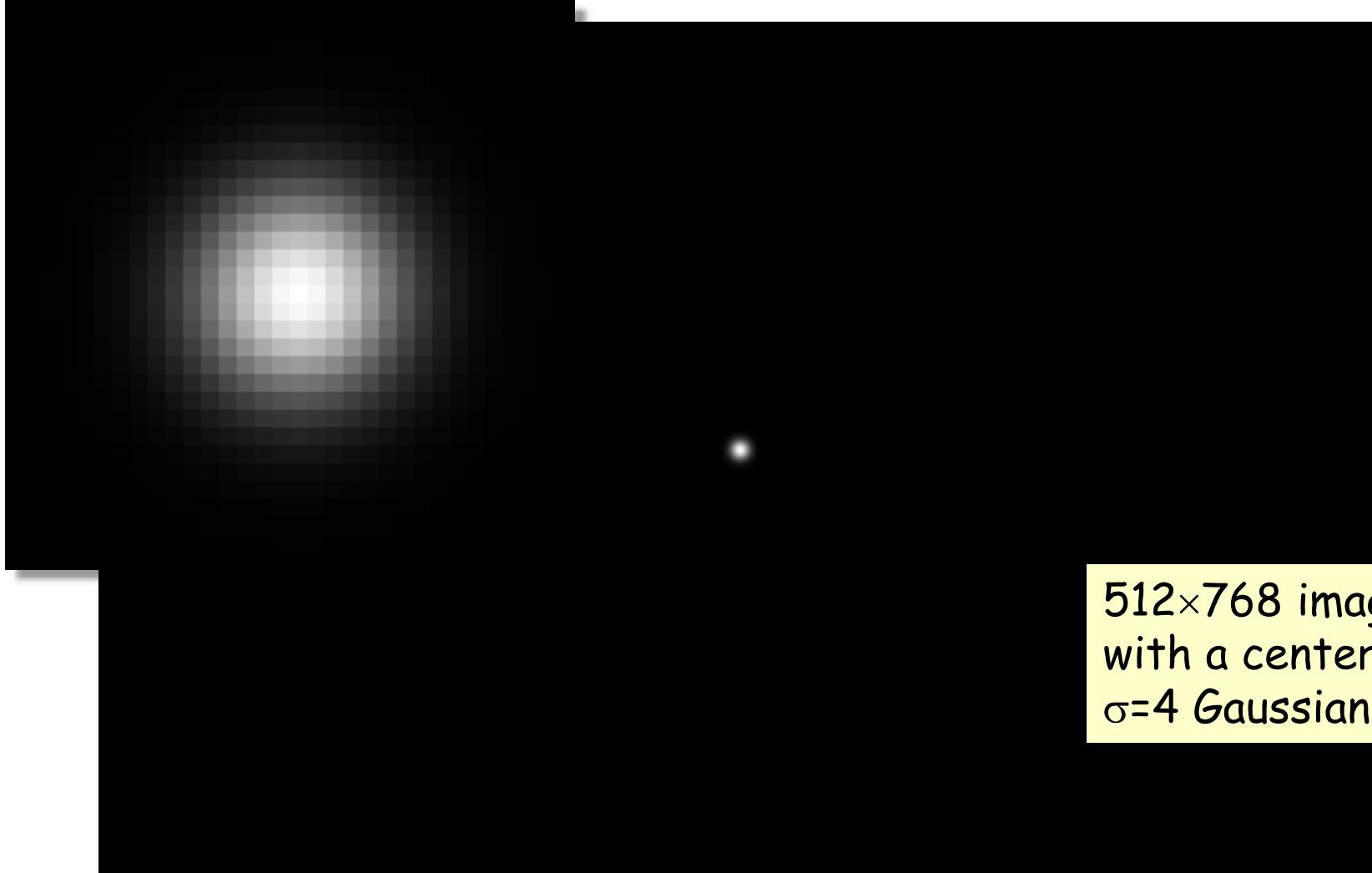
Simply copy the FT of the downsampled image into the center of a complex-valued image of all zeros that has twice the dimensions of the downsampled image. Then take the inverse FT.

This follows from the FT that results from multiplying the image with a “checkerboard” sampling grid. The following example with a Gaussian distribution demonstrates this.



Reconstruction

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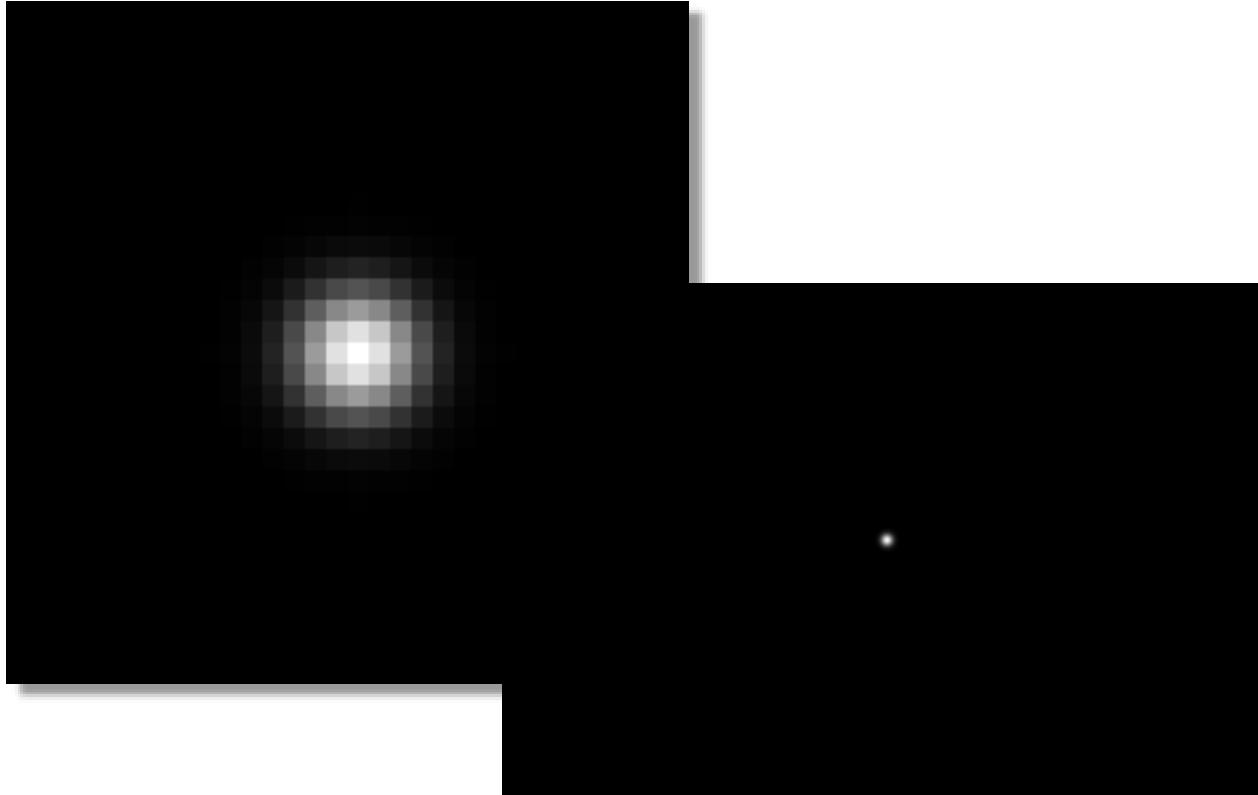


512×768 image
with a centered,
 $\sigma=4$ Gaussian.



Reconstruction

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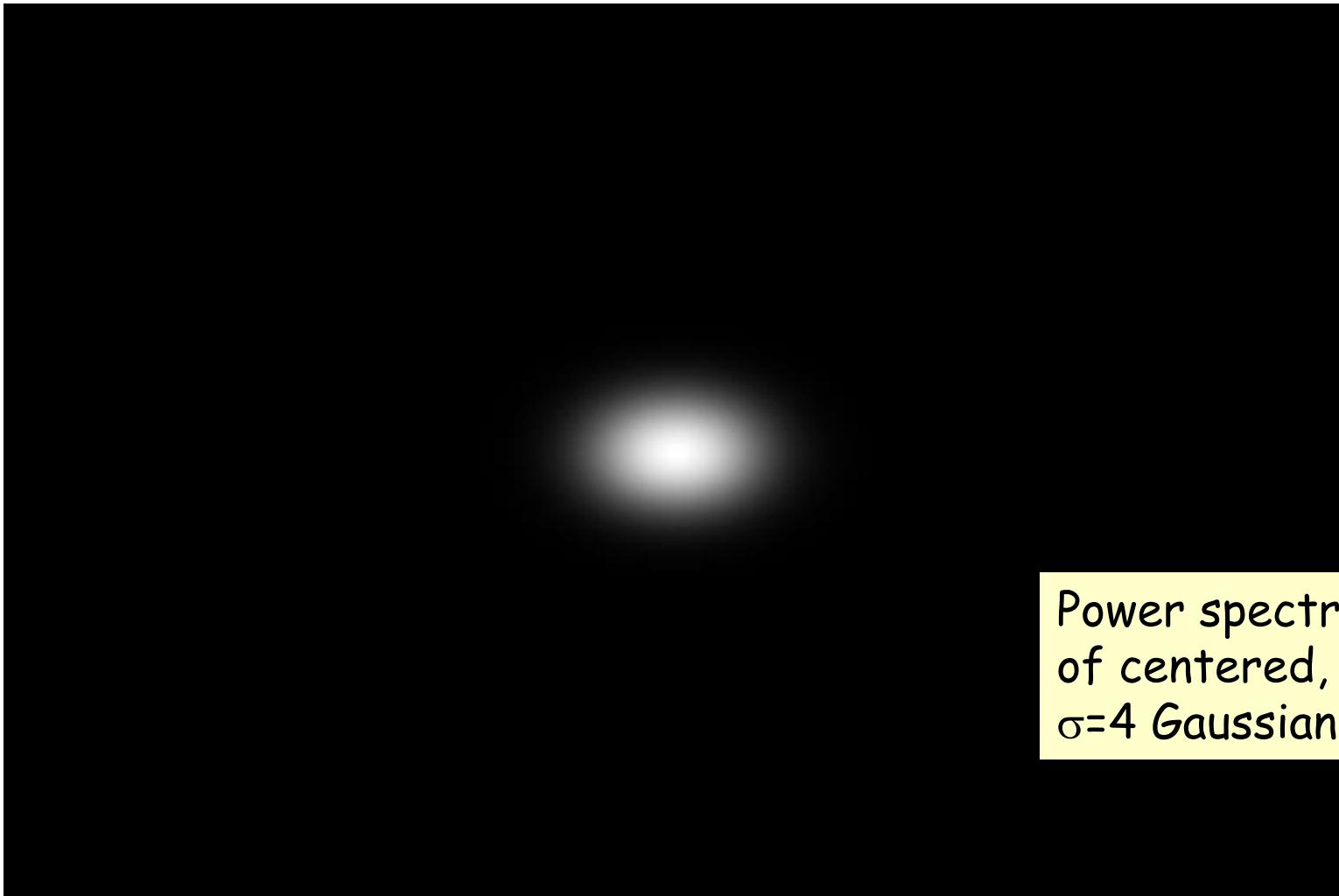


256×384 image
of downsampled
(1:2:512,1:2:768)
 $\sigma=4$ Gaussian.



Reconstruction

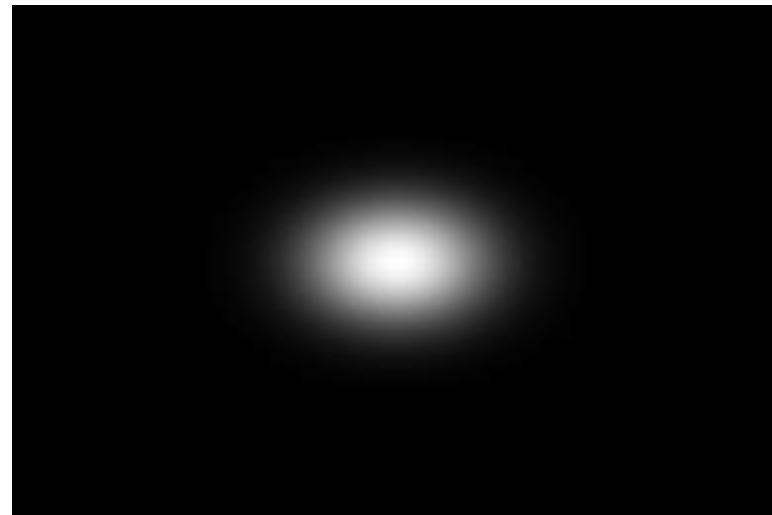
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Reconstruction

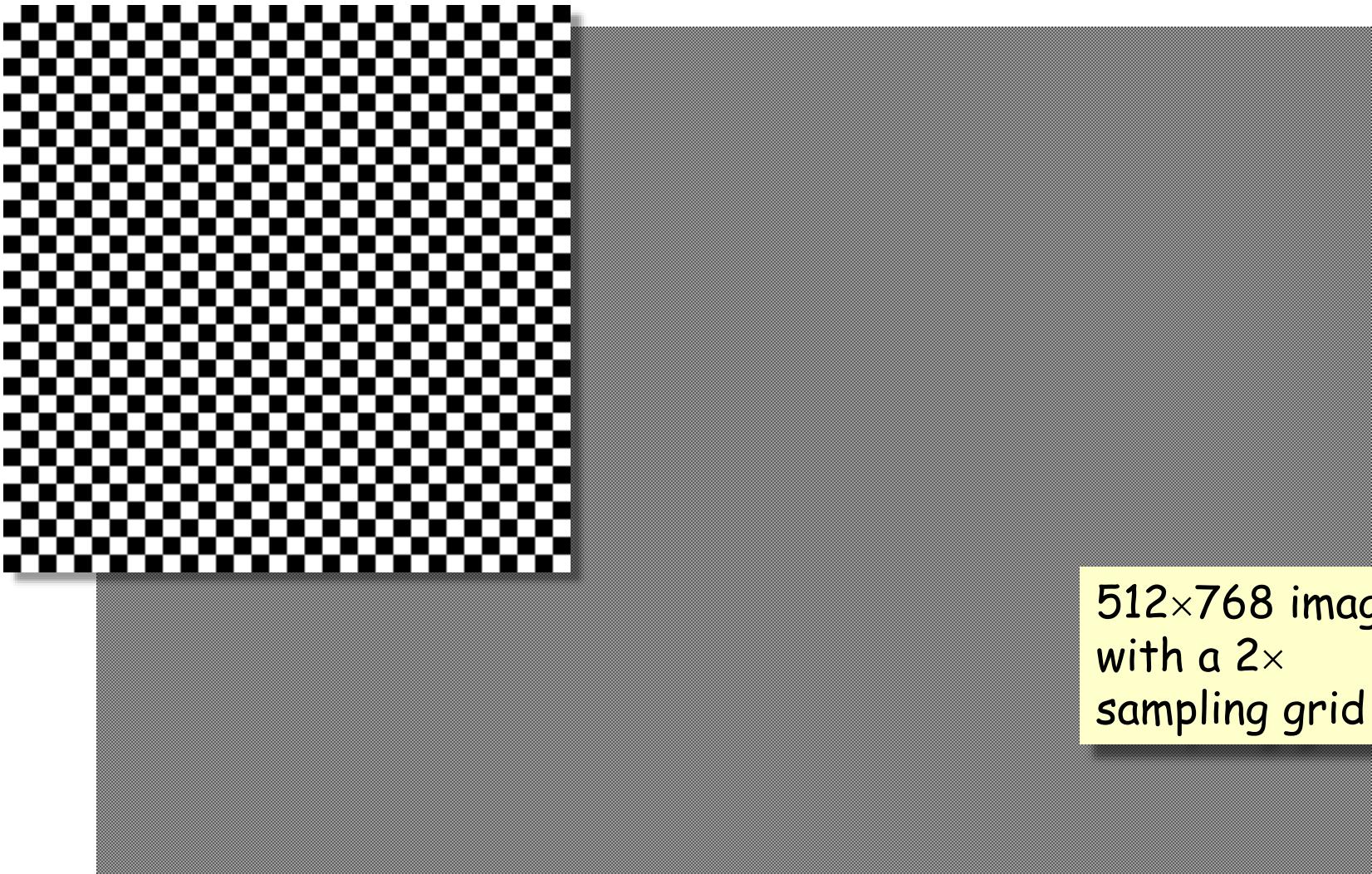
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Power spectrum
of downsampled
 $\sigma=4$ Gaussian.

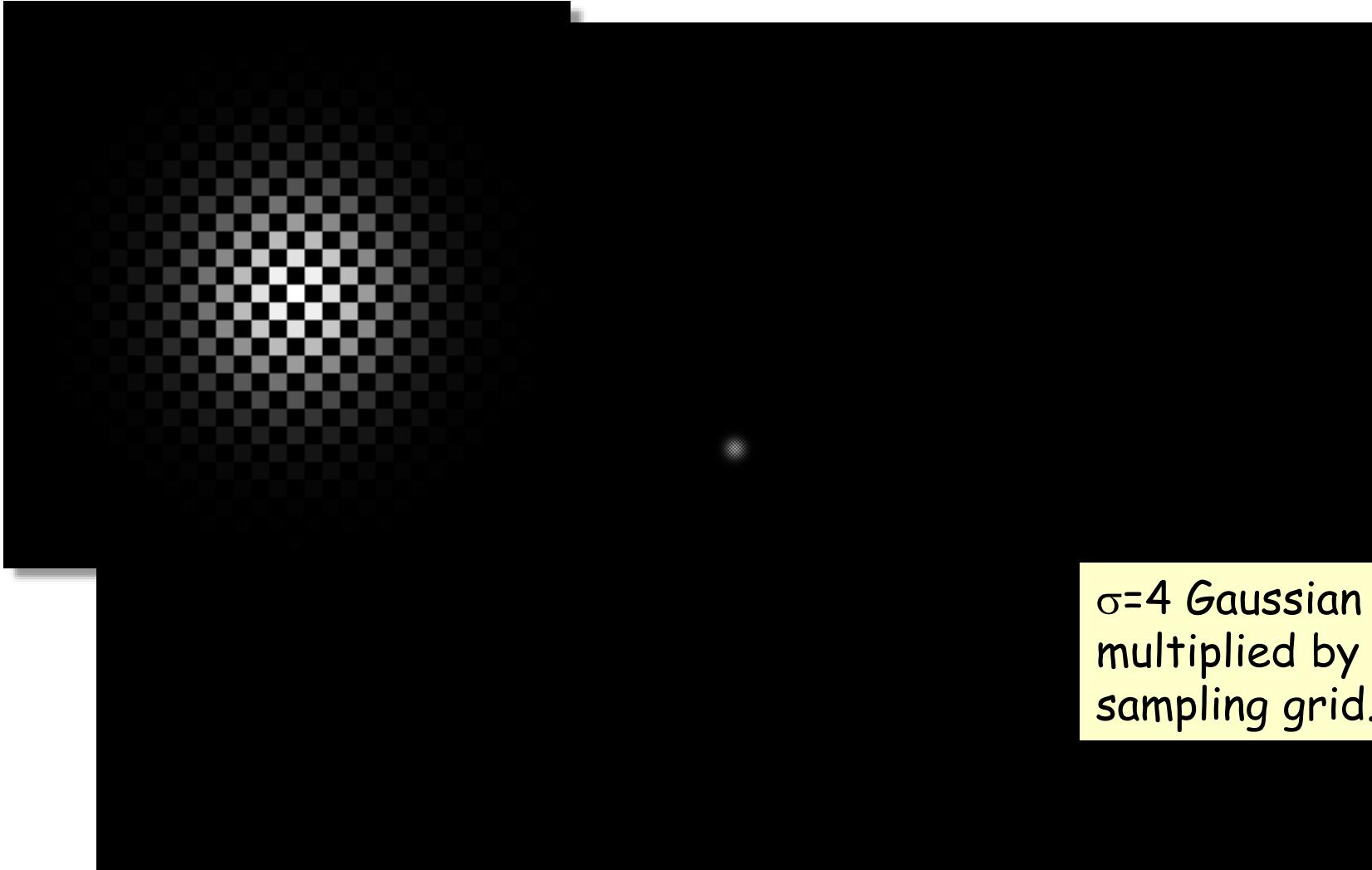


Reconstruction





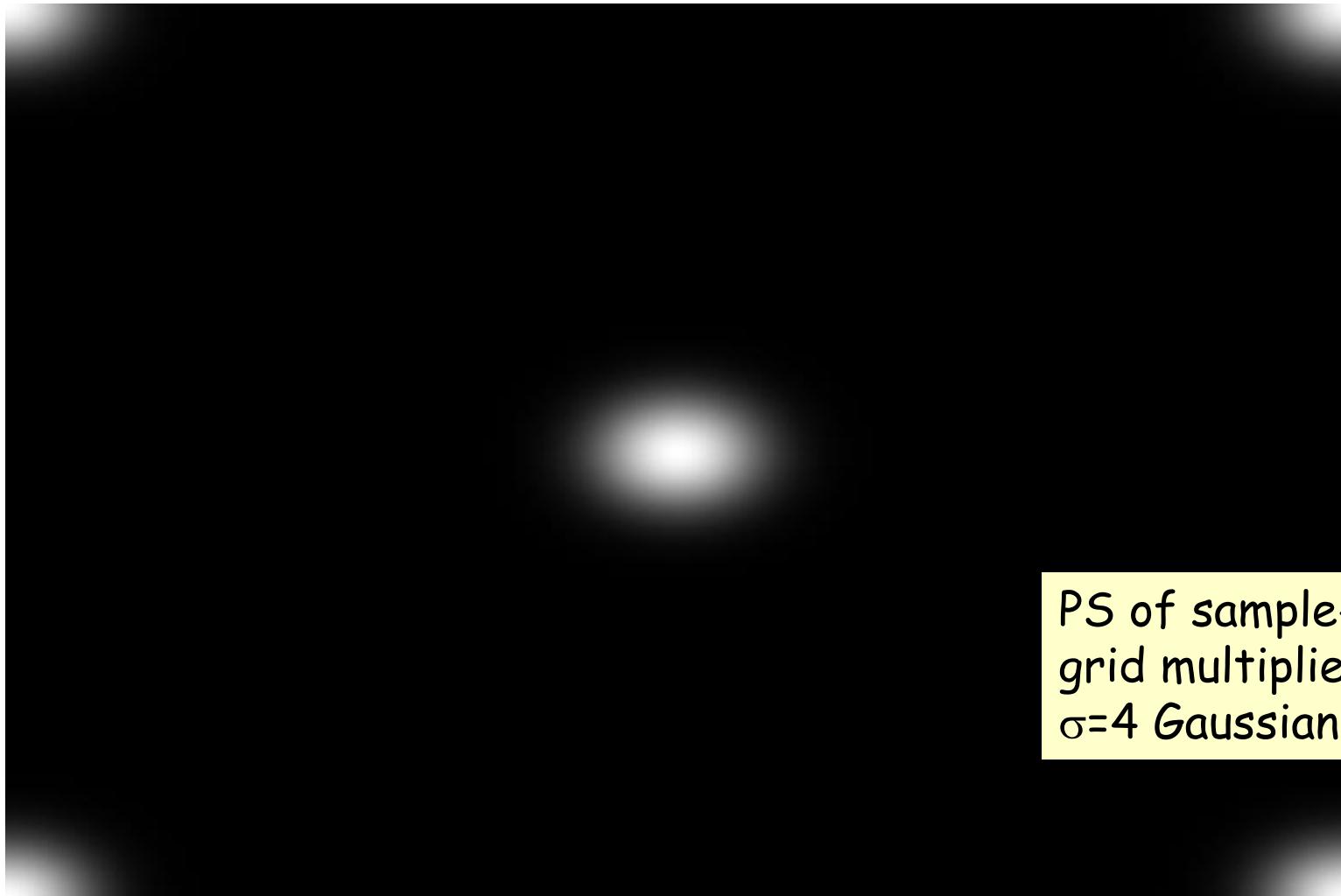
Reconstruction





Reconstruction

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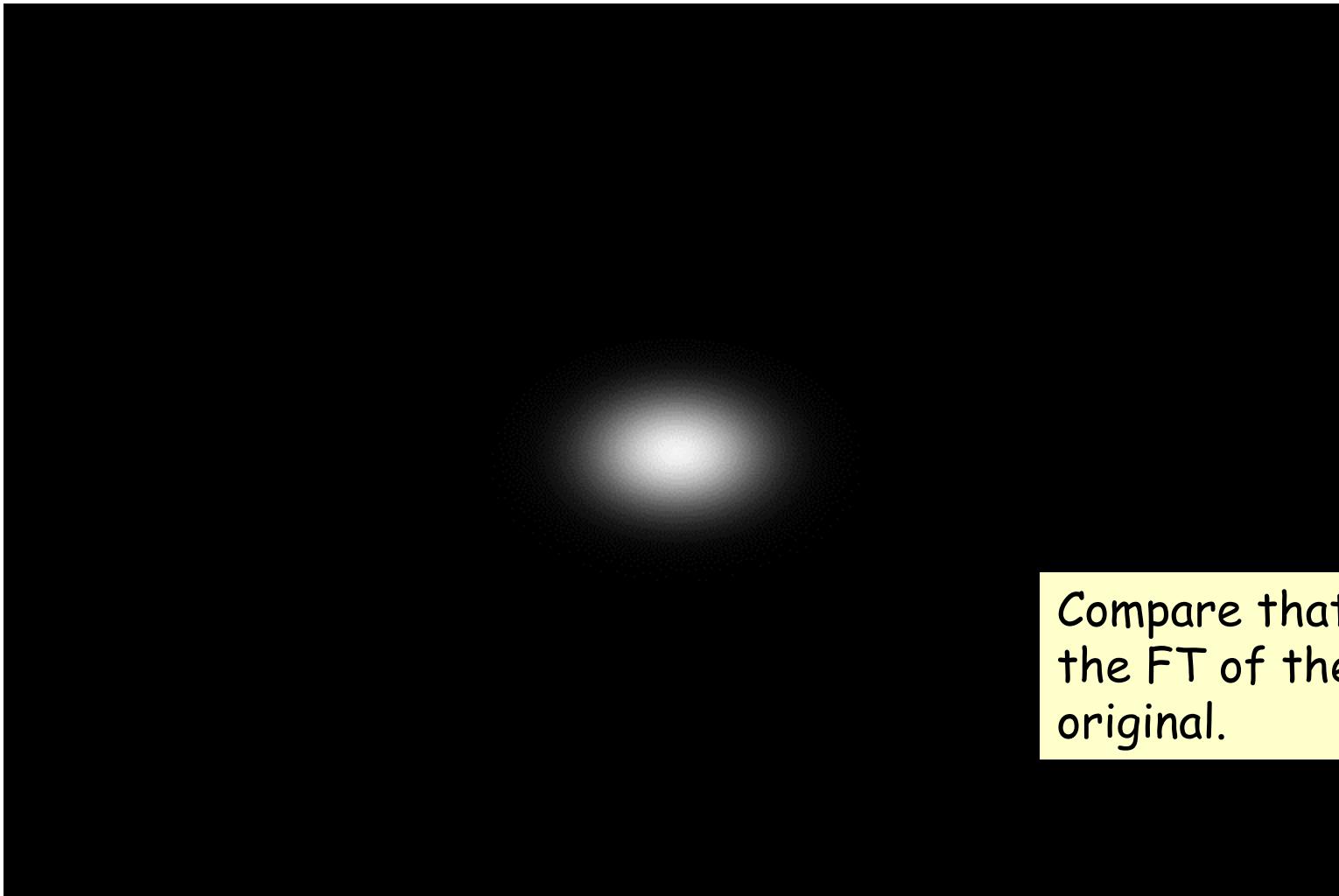


PS of sample-grid multiplied,
 $\sigma=4$ Gaussian.



Reconstruction

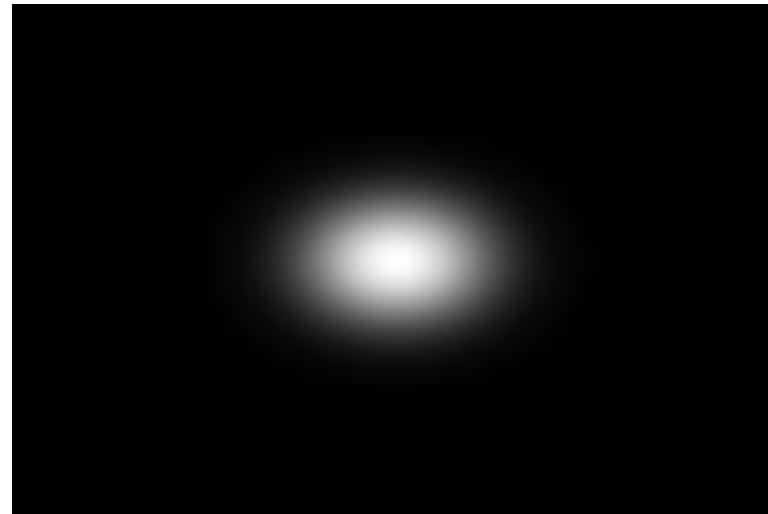
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Compare that to
the FT of the
original.



Reconstruction

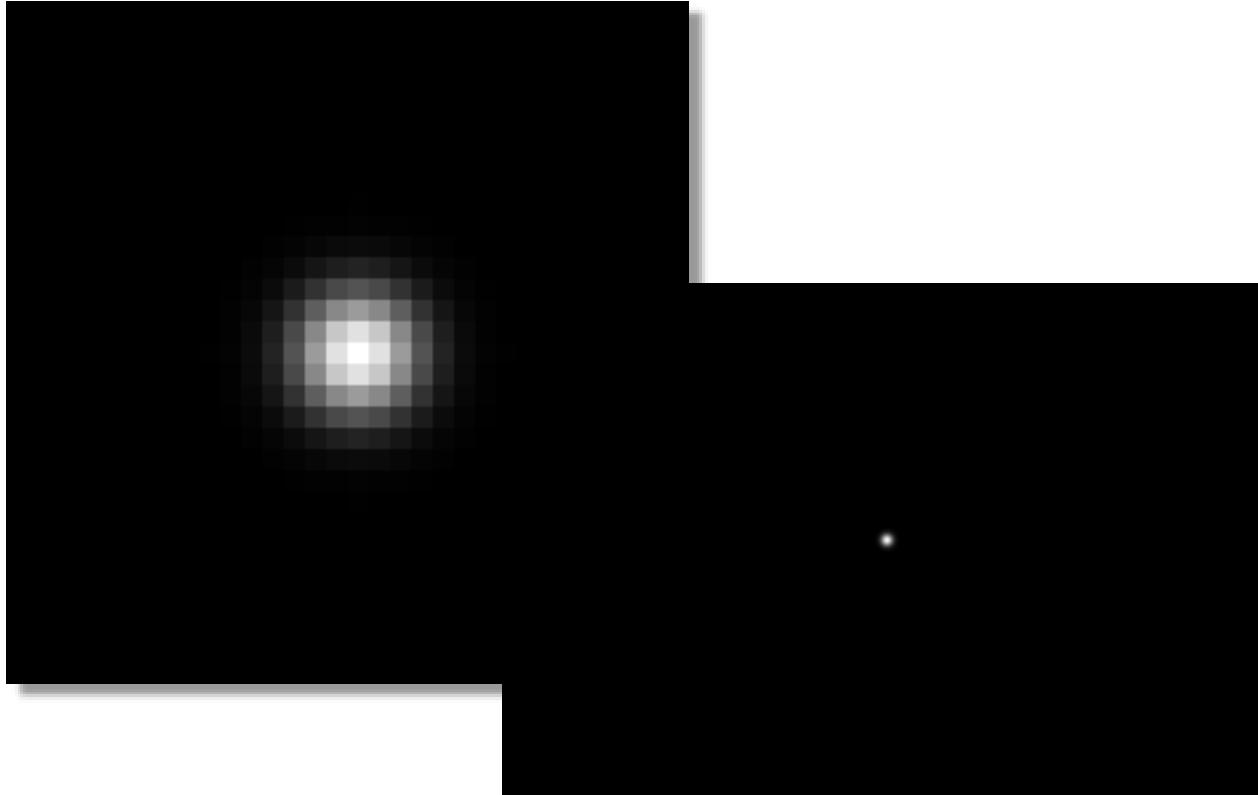


PS from
cropped FT of
sample-grid
multiplied $\sigma=4$
Gaussian.



Reconstruction

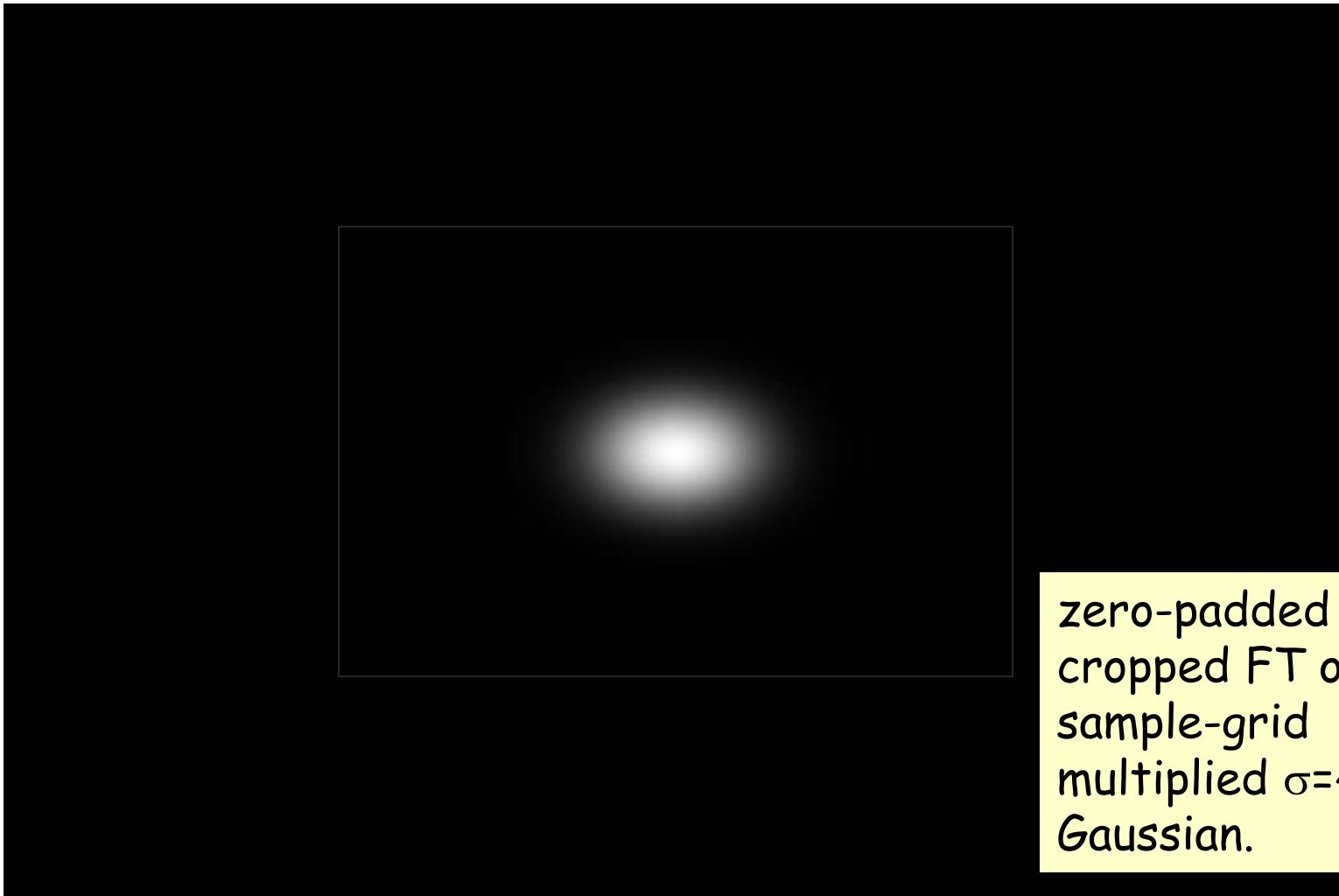
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IFT of cropped
sample-grid
mpy'd FT of $\sigma=4$
Gaussian.



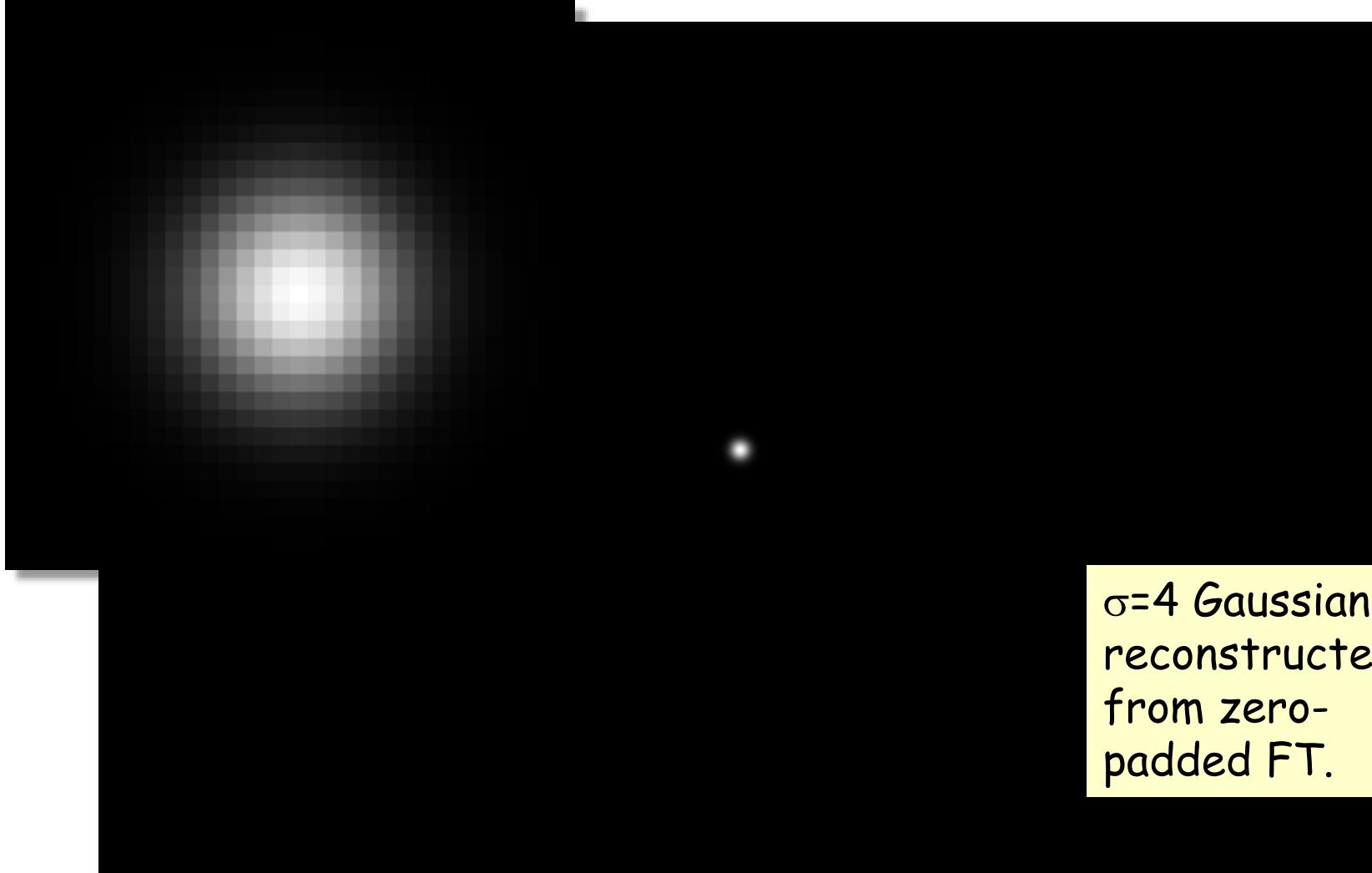
Reconstruction





Reconstruction

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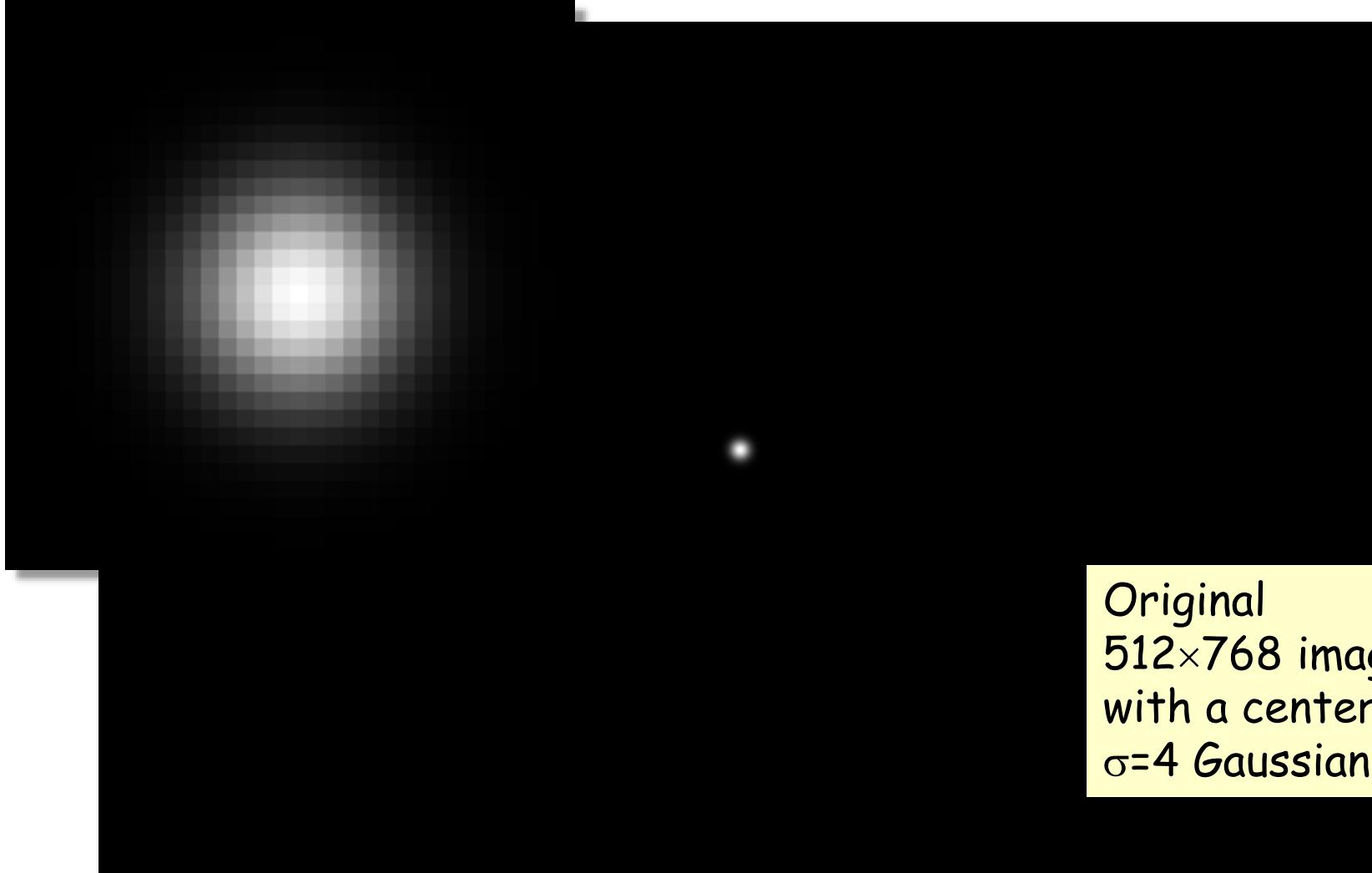


$\sigma=4$ Gaussian
reconstructed
from zero-
padded FT.



Reconstruction

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Original
512×768 image
with a centered,
 $\sigma=4$ Gaussian.