



EECE 4353 Image Processing

Lecture Notes: Reduction of Correlated Noise

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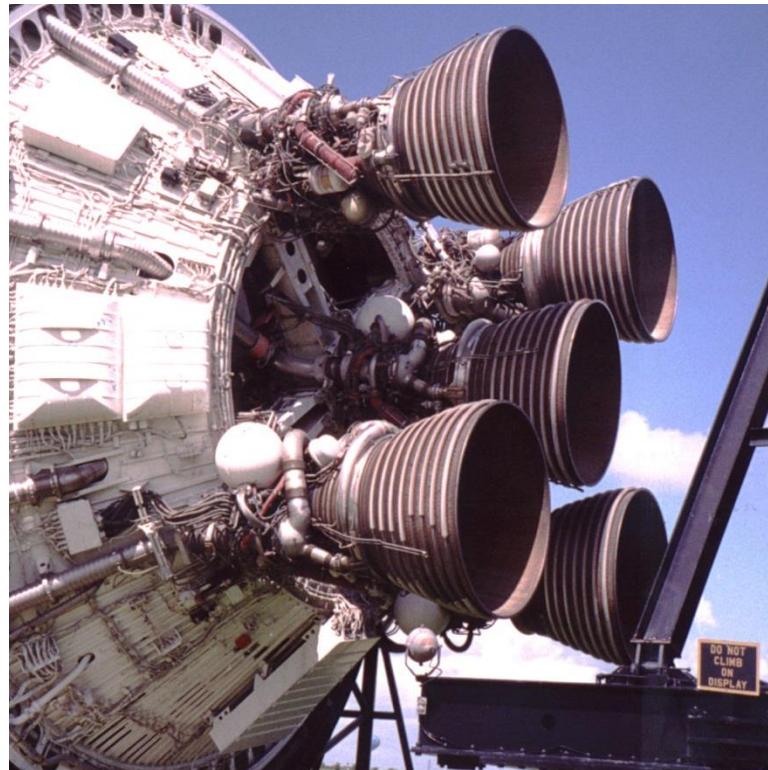
Fall Semester 2016





Pratt & Whitney Rocketdyne J-2 rocket engines
on Apollo 18's Saturn V second stage.

Periodic Noise



original image



image + noise



Pratt & Whitney Rocketdyne J-2 rocket engines
on Apollo 18's Saturn V second stage.

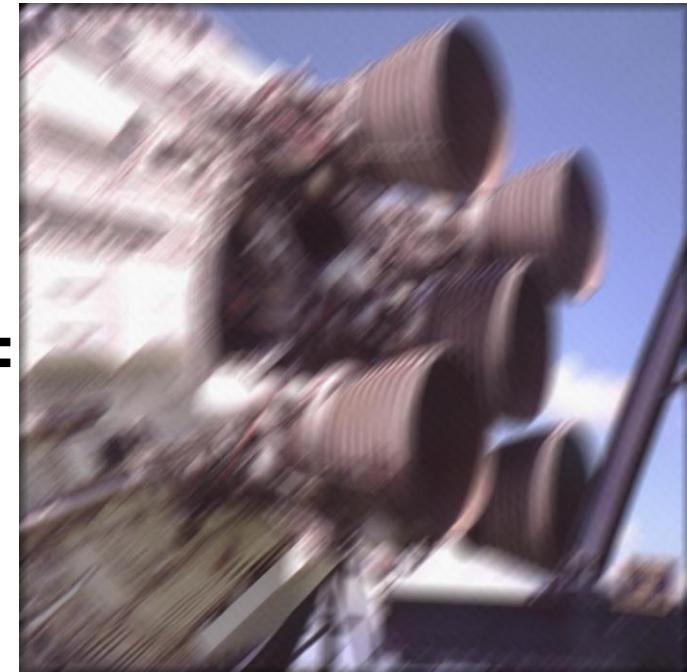
Noise Reduction through Directional Blurring



image + noise

A square grid of 16 small squares. A single diagonal line runs from the bottom-left square to the top-right square, representing the identity matrix.

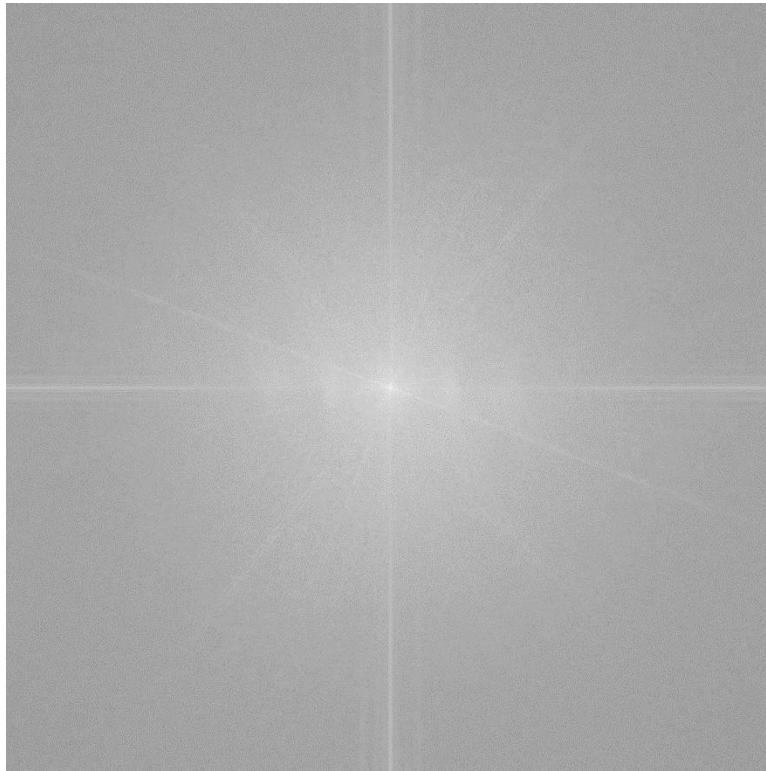
diagonal convolution mask



blurred image



Power Spectrum of Image with Periodic Noise



original image

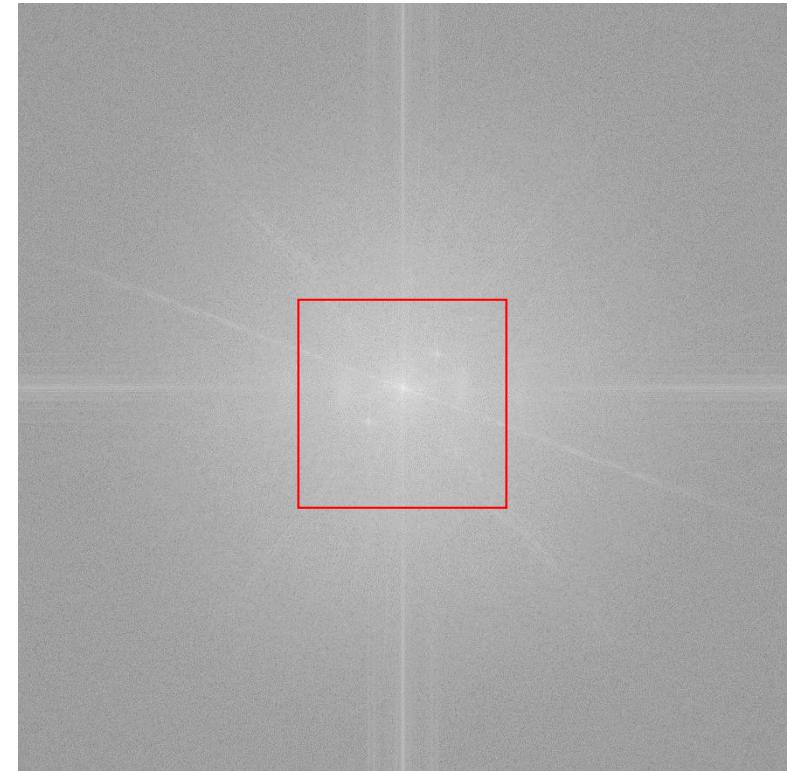
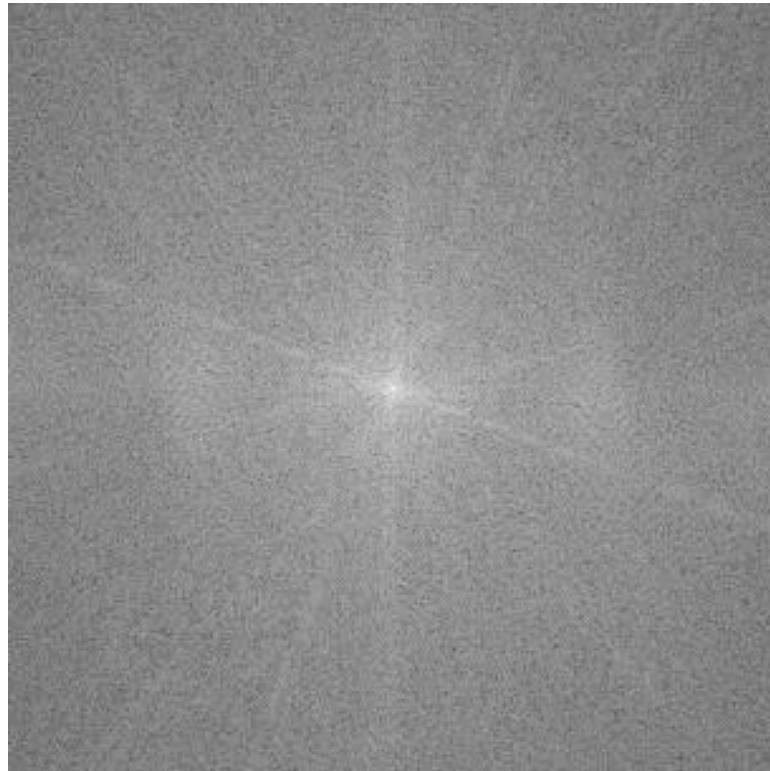


image + noise



Low Frequency Region



original image

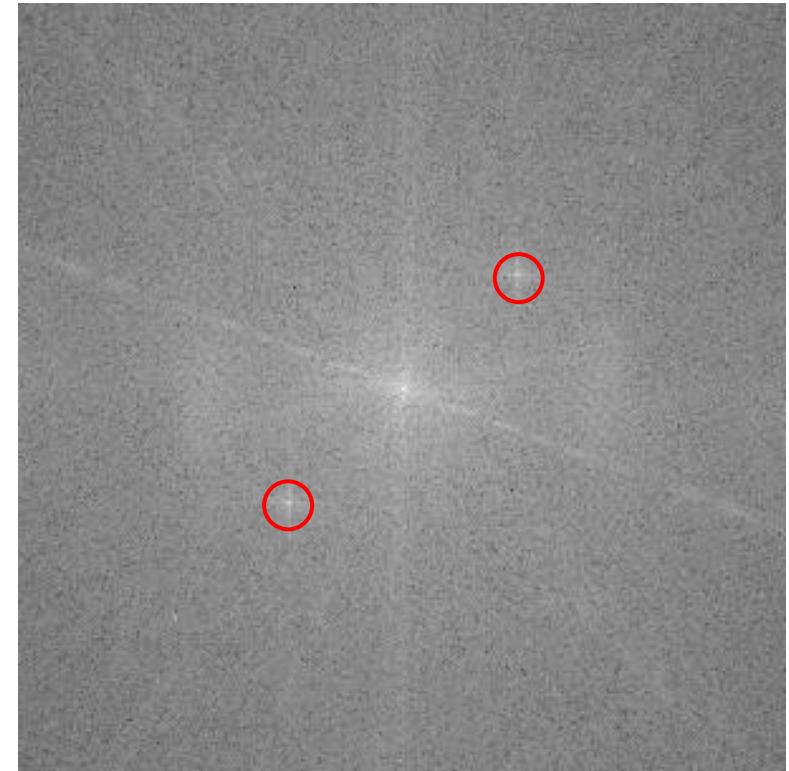
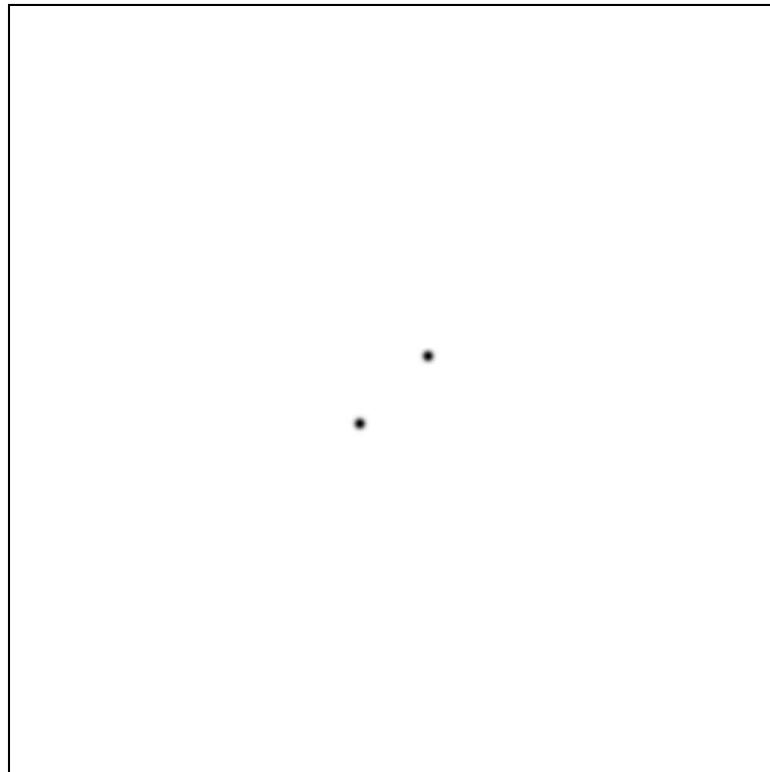


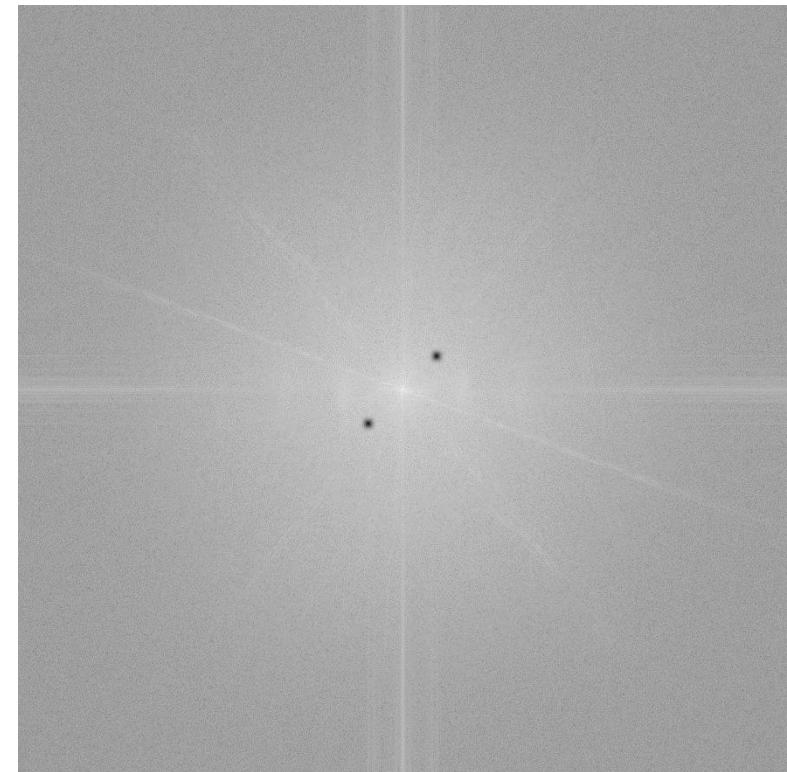
image + noise



Noise Reduction through Notch Filtering



noise mask

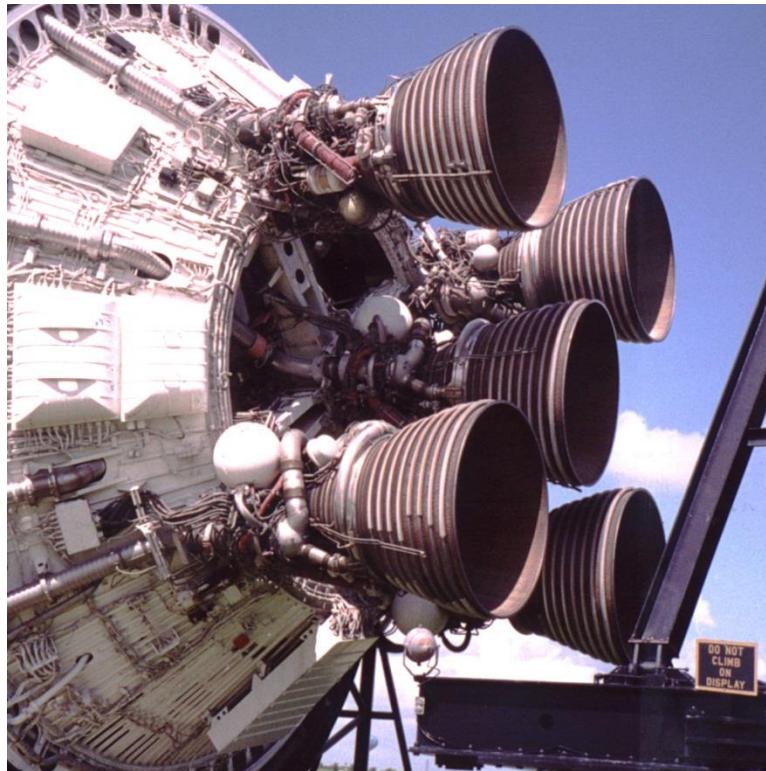


masked power spectrum

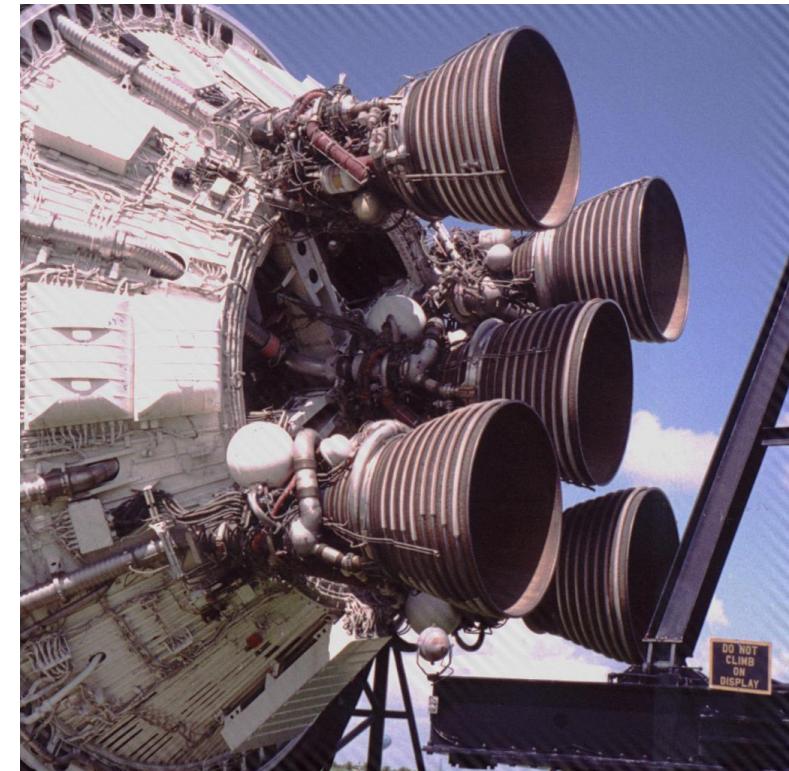


Pratt & Whitney Rocketdyne J-2 rocket engines
on Apollo 18's Saturn V second stage.

Inverse of Masked Fourier Transform



original image



noise reduced image

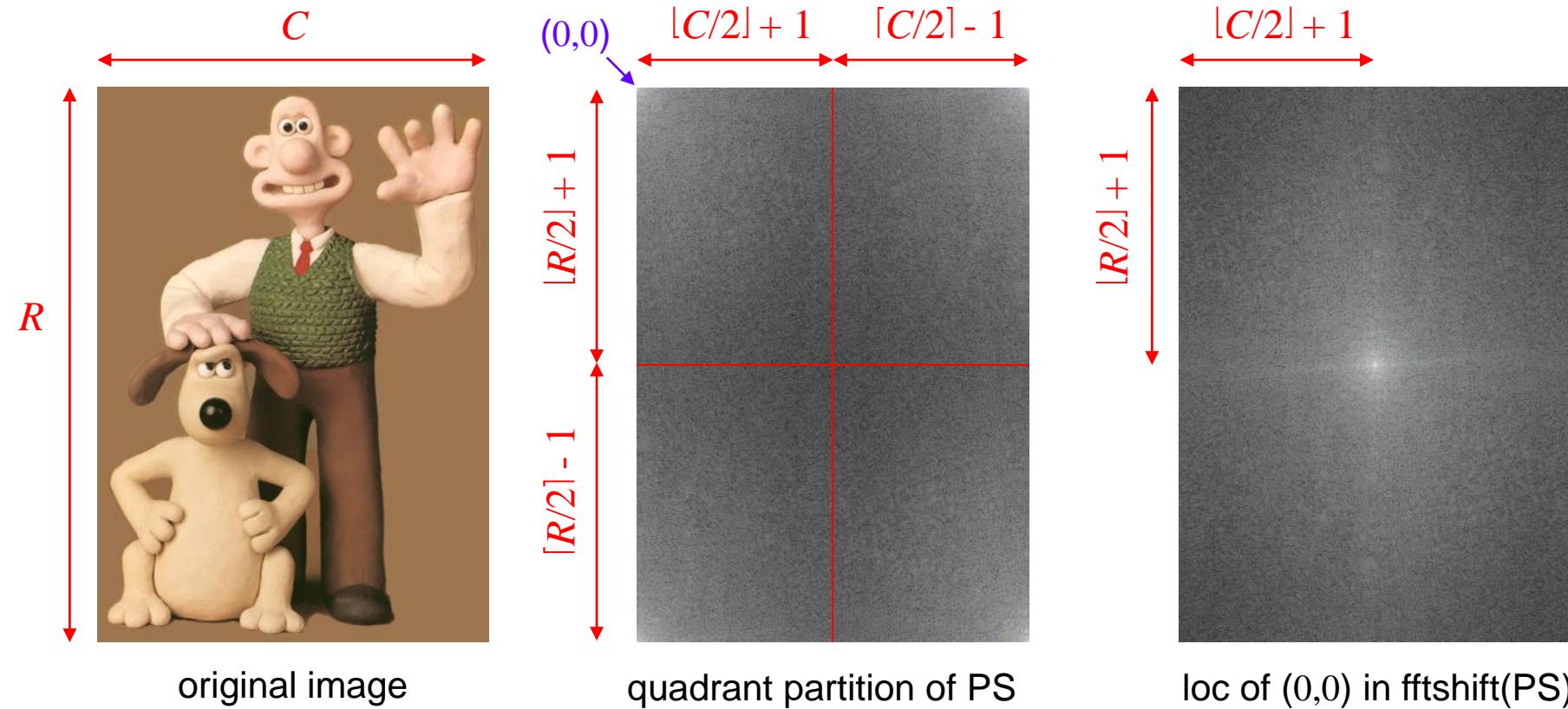


Notch filter reduction of periodic noise

1. Take the FFT of the input image. $\mathbf{F} = \text{fft2}(\mathbf{I})$.
 2. Display the log power spectrum of the `fftshift-ed` \mathbf{F} .
 3. Find the locations, \mathbf{x}_i , of the spikes that correspond to the periodic distortion.
 4. Create a 1-band image, \mathbf{M} , of class `double` the same size as \mathbf{I} .
 5. Set all \mathbf{M} 's pixels to 1.0.
 6. Let $\mathcal{N}(\mathbf{x}_i)$ be a neighborhood of \mathbf{x}_i with area sufficient to cover the spike at \mathbf{x}_i .
 7. For each \mathbf{x}_i do: for each $\mathbf{y}_j \in \mathcal{N}(\mathbf{x}_i)$, set $\mathbf{M}(\mathbf{y}_j) = 0$.
 8. Blur \mathbf{M} with a Gaussian whose σ is smaller than $\frac{1}{2}$ the radius of $\mathcal{N}(\mathbf{x}_i)$.
 9. Take the `ifftshift` of \mathbf{M} .
 10. For each band, k , of the image let $\mathbf{G}_k = \mathbf{F}_k .* \mathbf{M}$.
 11. Then the noise-reduced image is: $\mathbf{J} = \text{real}(\text{ifft2}(\mathbf{G}))$.
-

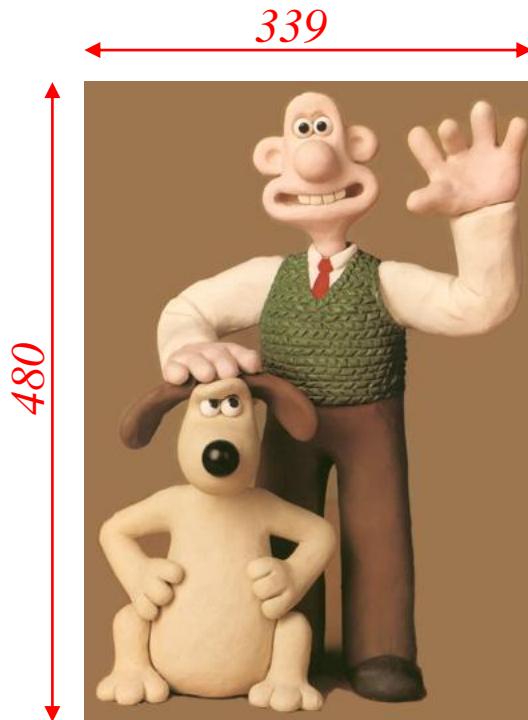


How to determine the frequency and period of a point in the log power spectrum of an image (ex.):

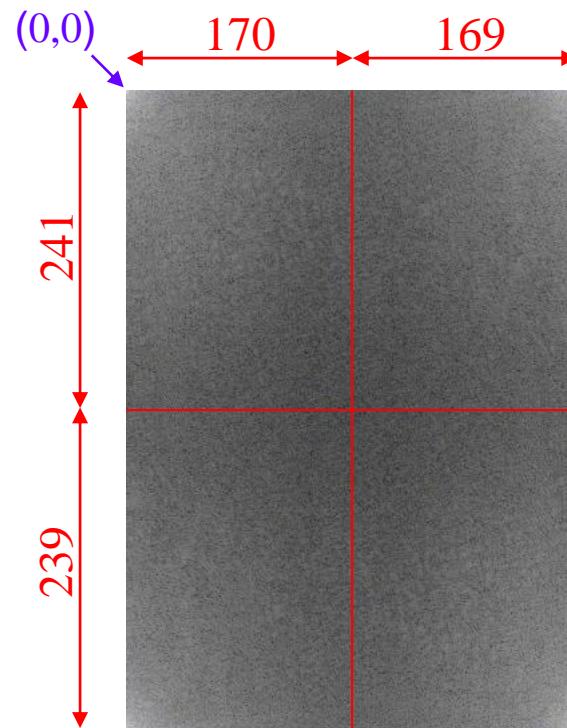




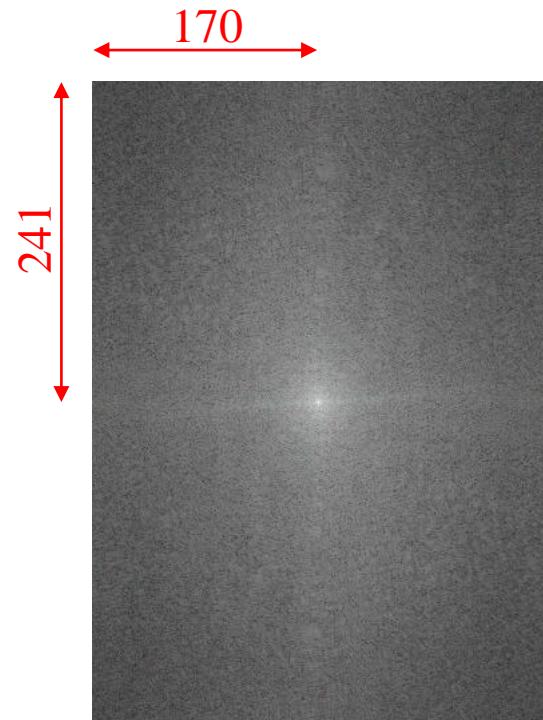
How to determine the frequency and period of a point in the log power spectrum of an image (ex.):



original image



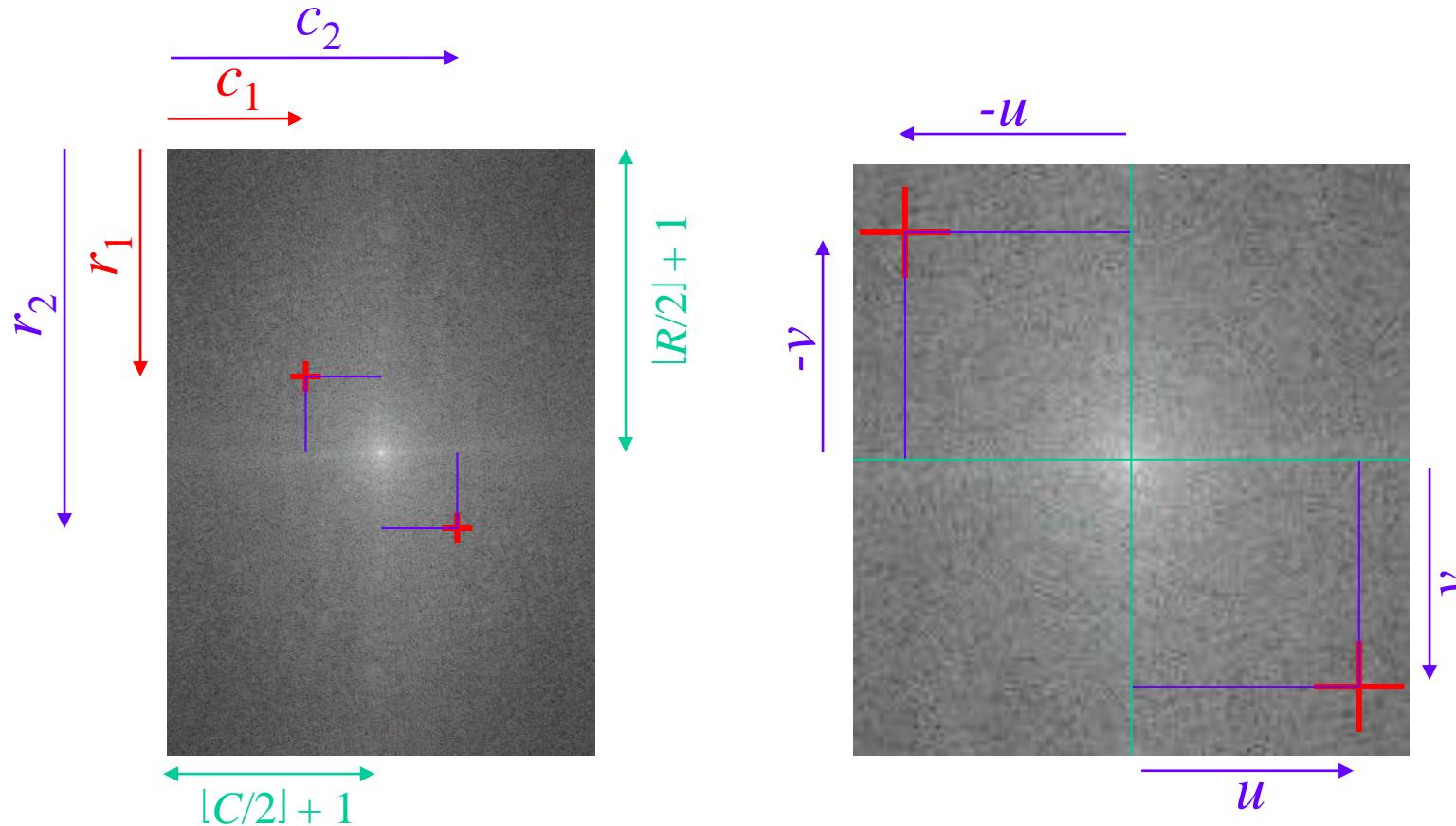
quadrant partition of PS



loc of $(0,0)$ in `fftshift(PS)`

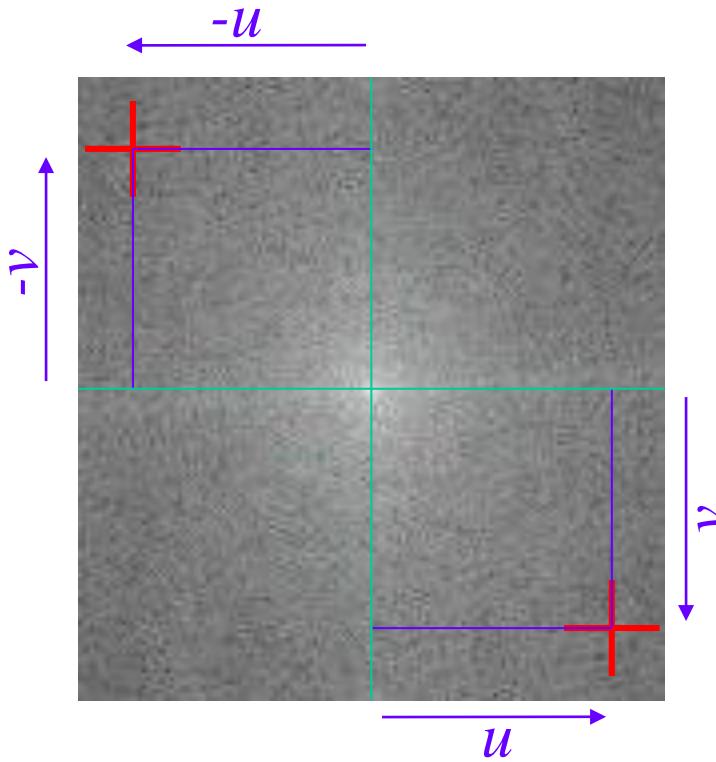


How to determine the frequency and period of a point in the log power spectrum of an image:





How to determine the frequency and period of a point in the log power spectrum of an image:



$$\begin{aligned}v &= r_2 - \lfloor R/2 \rfloor - 1 \\-v &= \lfloor R/2 \rfloor + 1 - r_1 \\u &= c_2 - \lfloor C/2 \rfloor - 1 \\-u &= \lfloor C/2 \rfloor + 1 - c_1\end{aligned}$$

$$\begin{aligned}\lambda_{wf} &= \sqrt{\left(\frac{C}{u}\right)^2 + \left(\frac{R}{v}\right)^2} \\\omega_{wf} &= \frac{1}{\lambda_{wf}} \\\theta_{wf} &= \tan^{-1}\left(\frac{vC}{uR}\right)\end{aligned}$$



Points on the Fourier Plane (of a Digital Image)

In the Fourier transform of an $R \times C$ digital image the wavelengths, λ_u and λ_v represent a fraction of the R and C values. That is,

$$\lambda_u = \frac{C}{u} \text{ and } \lambda_v = \frac{R}{v} \text{ pixels.}$$

The wavefront direction is given by

$$\theta_{wf} = \tan^{-1} \left(\frac{vC}{uR} \right),$$

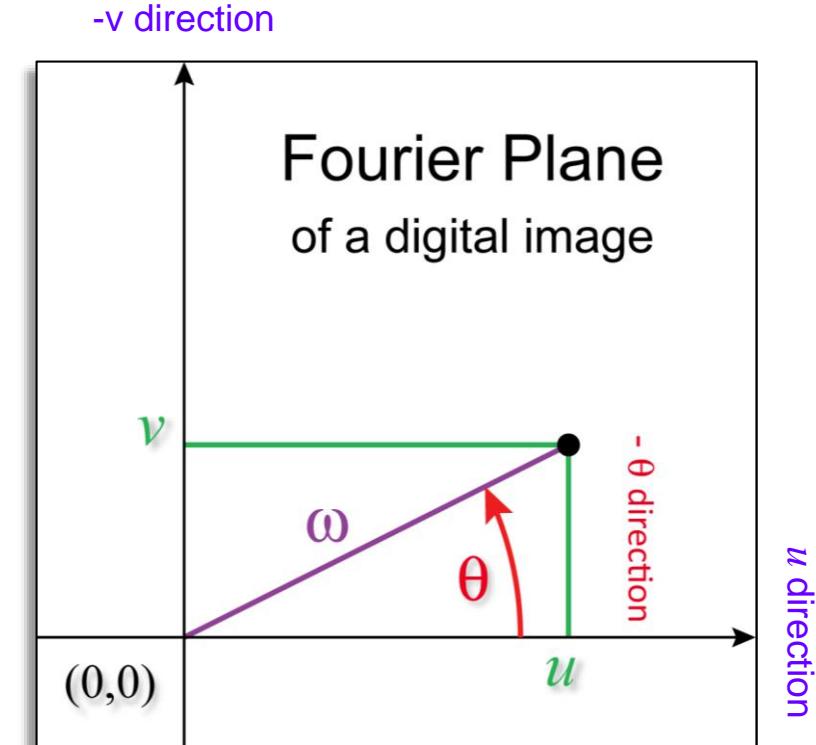
and the wavelength is

$$\lambda_{wf} = \sqrt{\left(\frac{C}{u}\right)^2 + \left(\frac{R}{v}\right)^2}.$$

The frequencies represent fractions of R & C ,

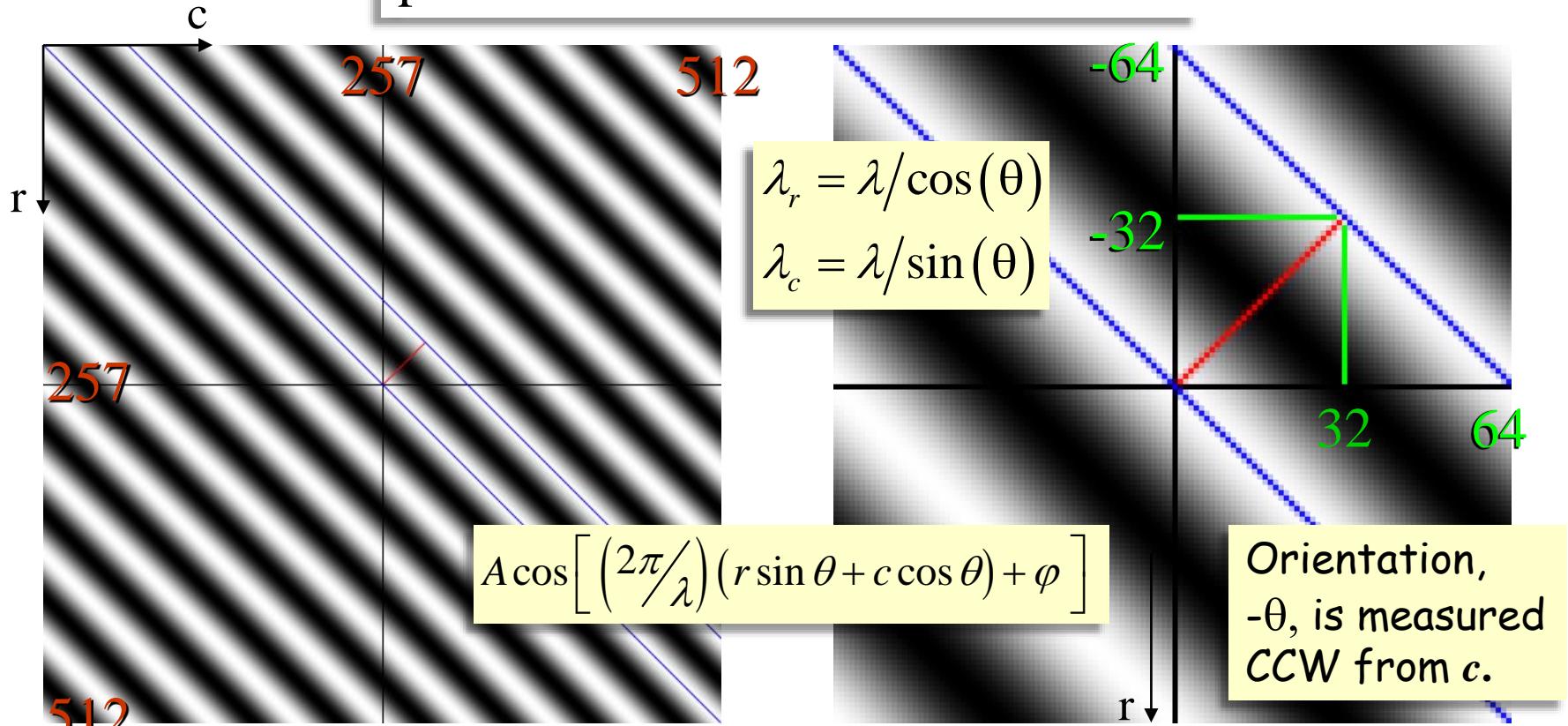
$$\omega_u = \frac{u}{C}, \quad \omega_v = \frac{v}{R}, \text{ and}$$

$$\omega_{wf} = 1 / \sqrt{\left(\frac{C}{u}\right)^2 + \left(\frac{R}{v}\right)^2} \text{ cycles.}$$





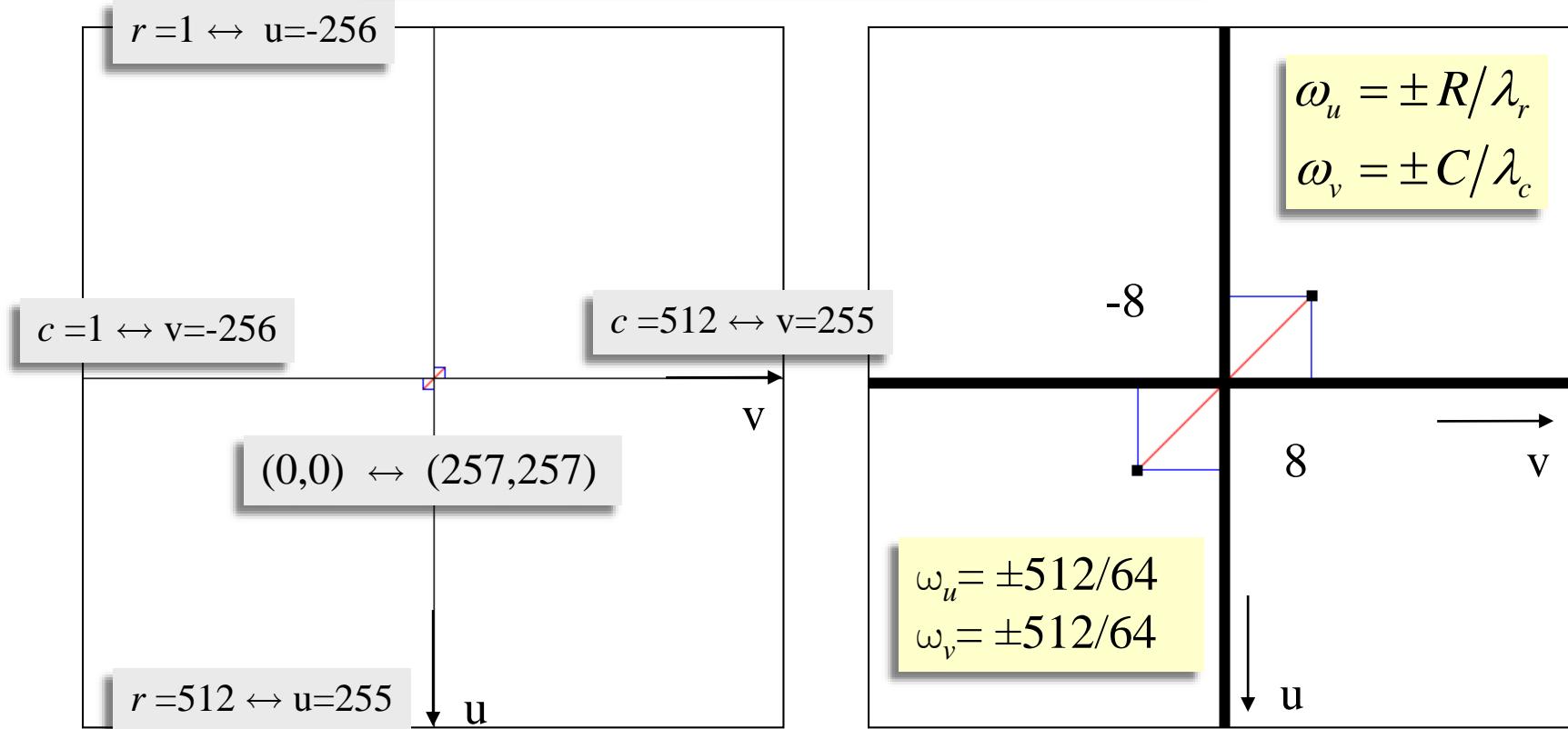
How to determine the frequency plane location of a sinusoid:



$$\text{cosine grating } \lambda = 32\sqrt{2}, \theta = -\pi/4 \Rightarrow \lambda_r = 64 = 512/v, \lambda_c = 64 = 512/v$$



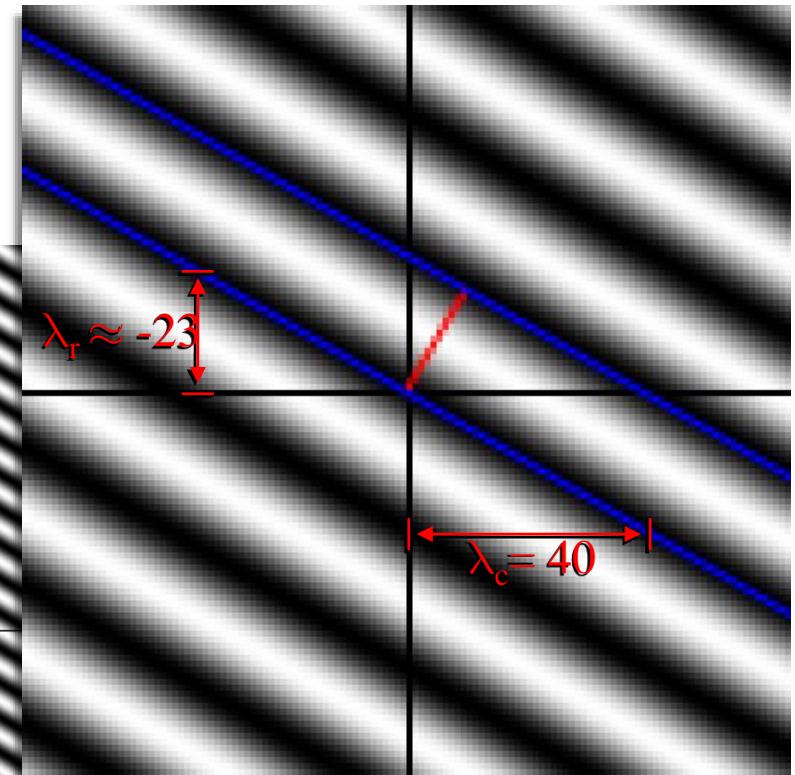
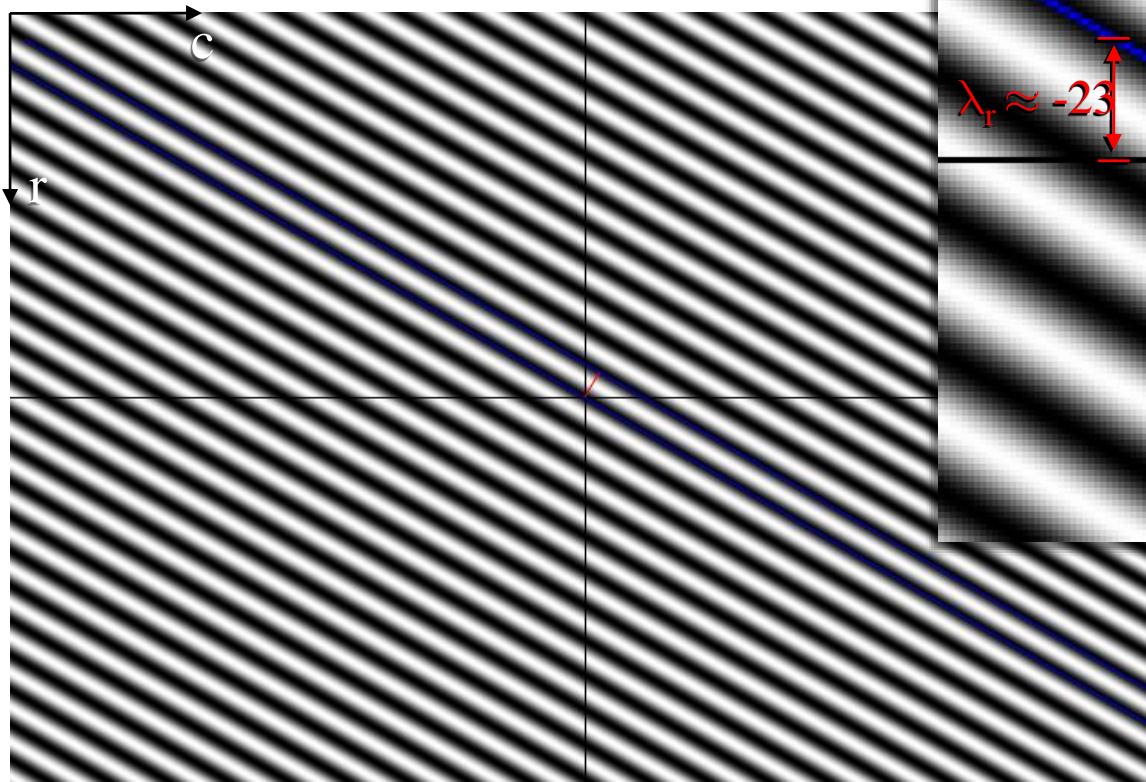
How to determine the frequency plane location of a sinusoid:



cosine grating $\omega = 8\sqrt{2}$, orientation = $3\pi/4 \Rightarrow \omega_u = \pm 8, \omega_v = \pm 8$



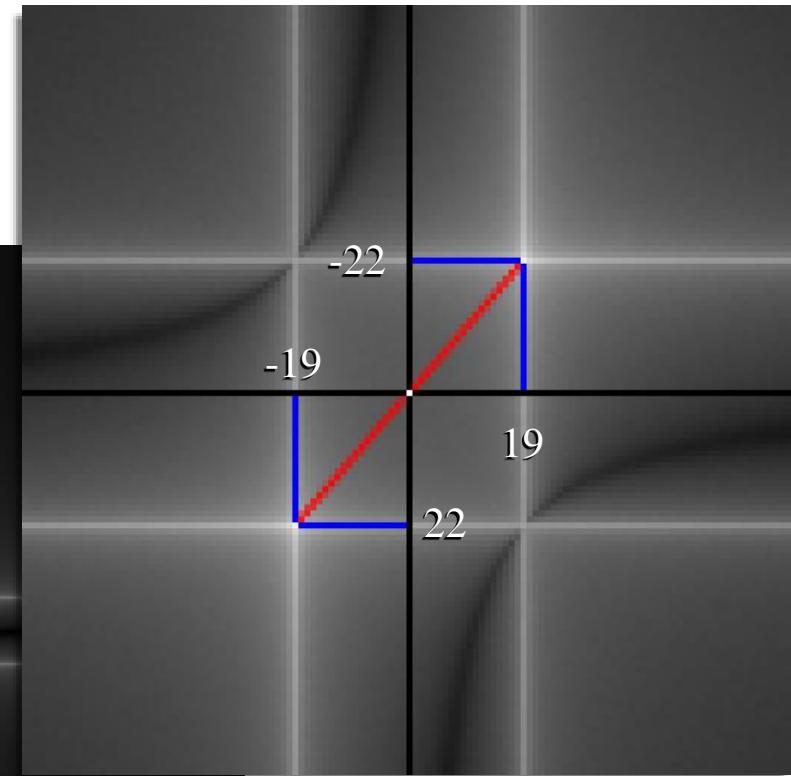
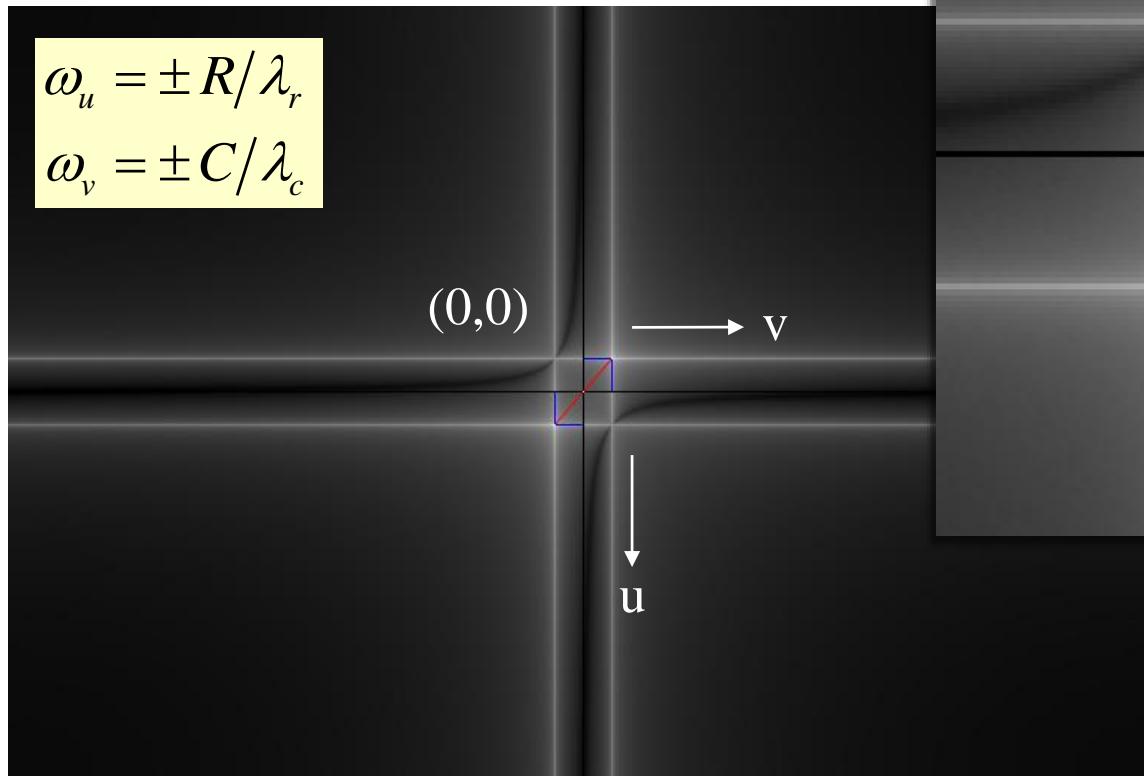
Sinusoidal grating image with an even number of rows and an odd number of columns.



$$\begin{aligned} & \text{512} \times 767 \text{ sine grating} \\ & \lambda = 20, \text{ orient.} = 5\pi/6 \\ & \Rightarrow \lambda_r = 20/\cos(5\pi/6) \approx -23, \\ & \lambda_c = 20/\sin(5\pi/6) = 40 \end{aligned}$$



FT of Sinusoidal Grating Image



752×937 sine grating
 $\lambda = 20$, orient. = $5\pi/6$
 $\Rightarrow \omega_u = \pm 512/\lambda_r \approx \pm 22$,
 $\omega_v = \pm 767/\lambda_c \approx \pm 19$



Processing Scanned Press-Printed Images

4-color printing:

1. A photograph or other color image is separated into four intensity band images: cyan, magenta, yellow, and black.
2. Each of these is multiplied by a halftone screen¹ – a dot mask with a unique orientation.
3. Each of the resulting four images shows a pattern of dots whose individual sizes indicate the amount of ink to be applied at each point.
4. The four images are printed, one atop the other, in the corresponding color.

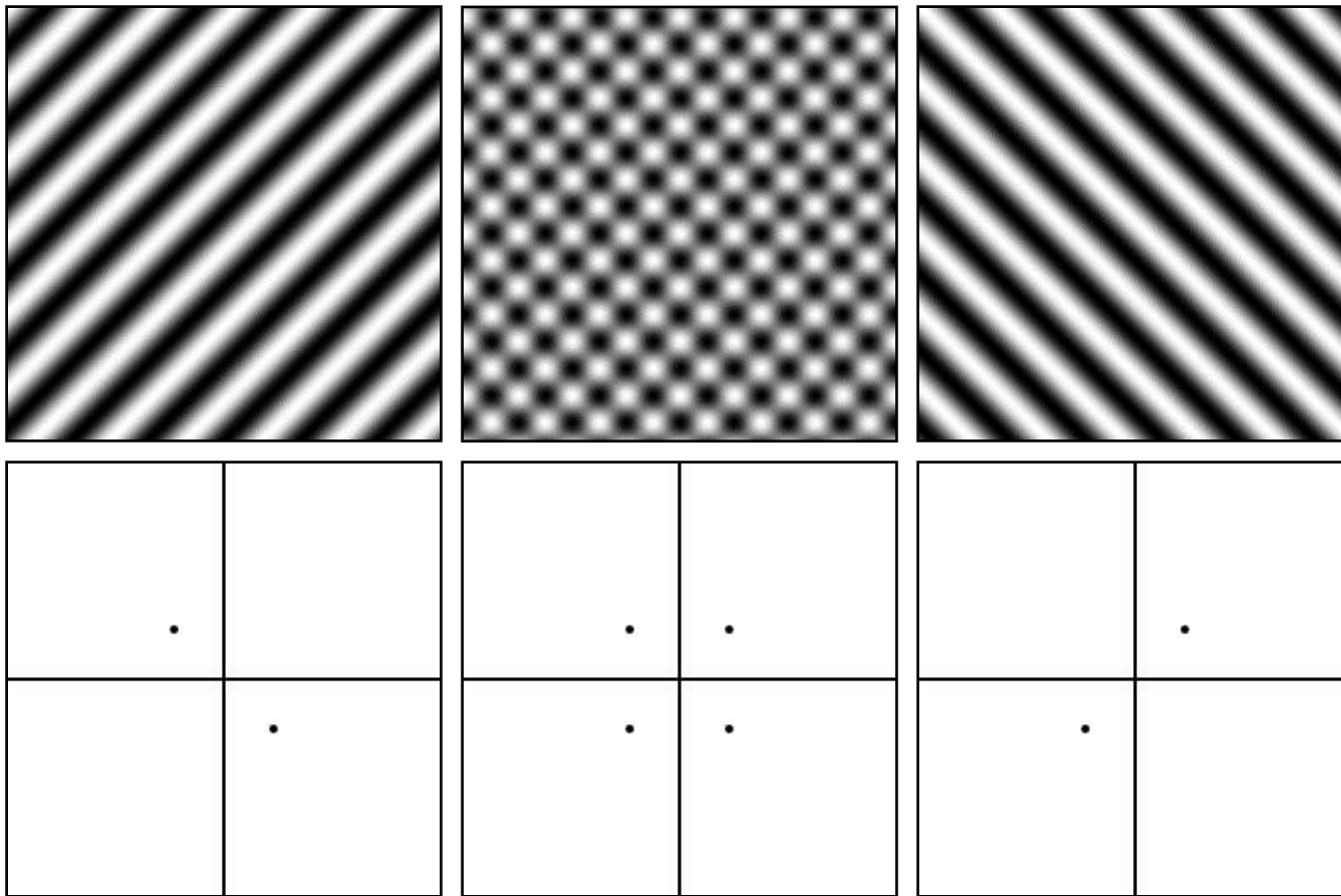
¹Merriam-Webster Dictionary: half·tone 2 (1): a photoengraving made from an image photographed through a screen and then etched so that the details of the image are reproduced in dots.



Halftone Screen (45°)

is the pointwise product of 2 sinusoidal gratings with perpendicular orientations.

space
frequency

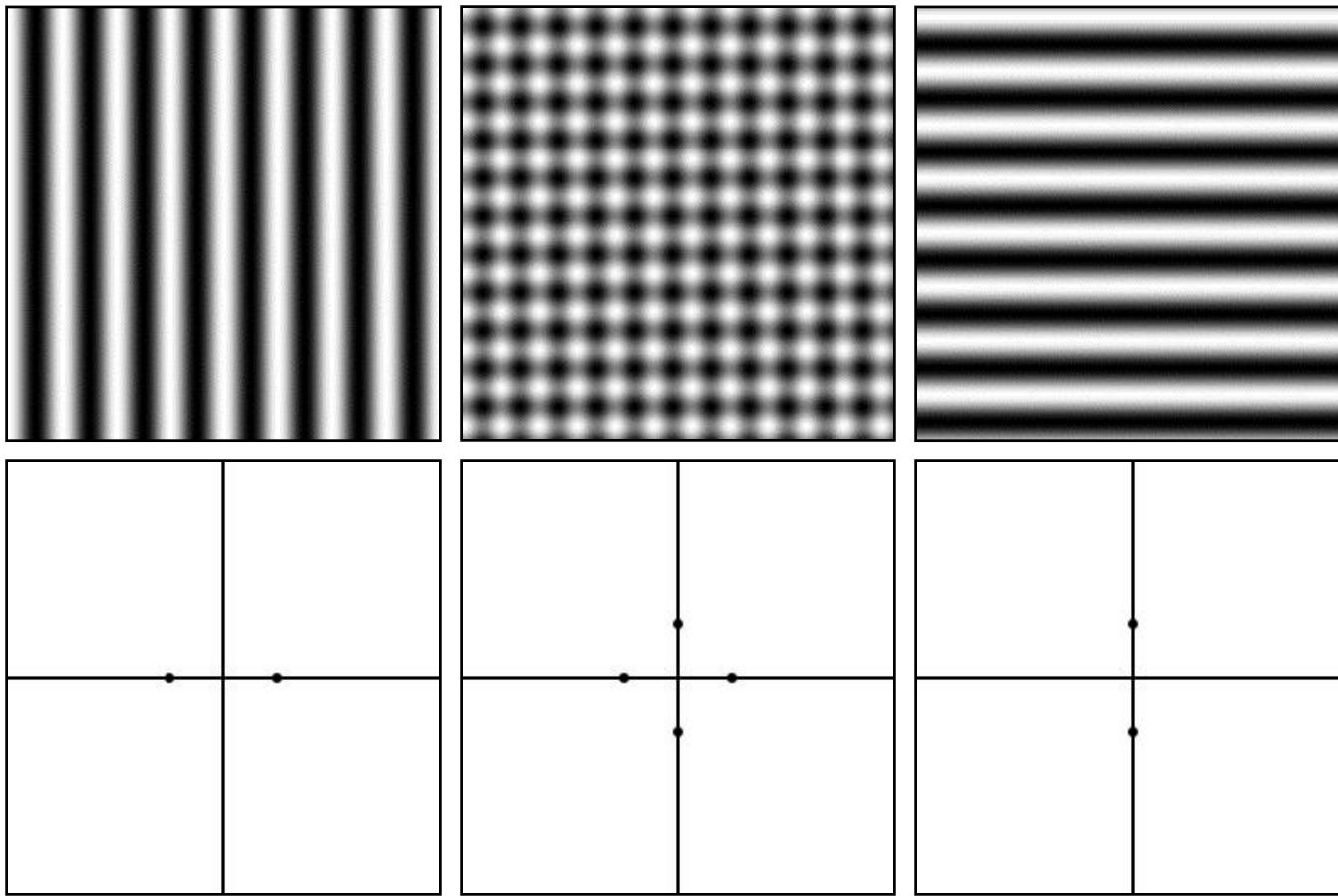




Halftone Screens (90°)

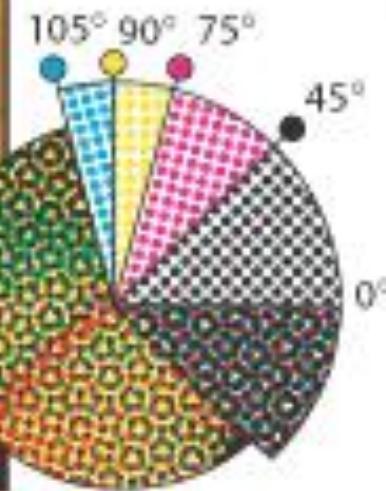
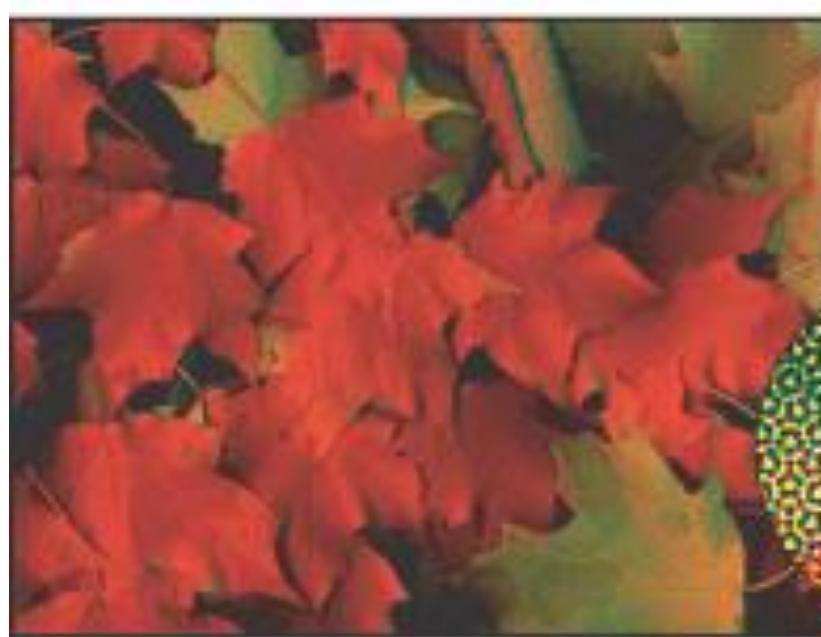
The orientation of the screen is the average of the grating orientations.

space
frequency





Standard Halftone Screen Angles



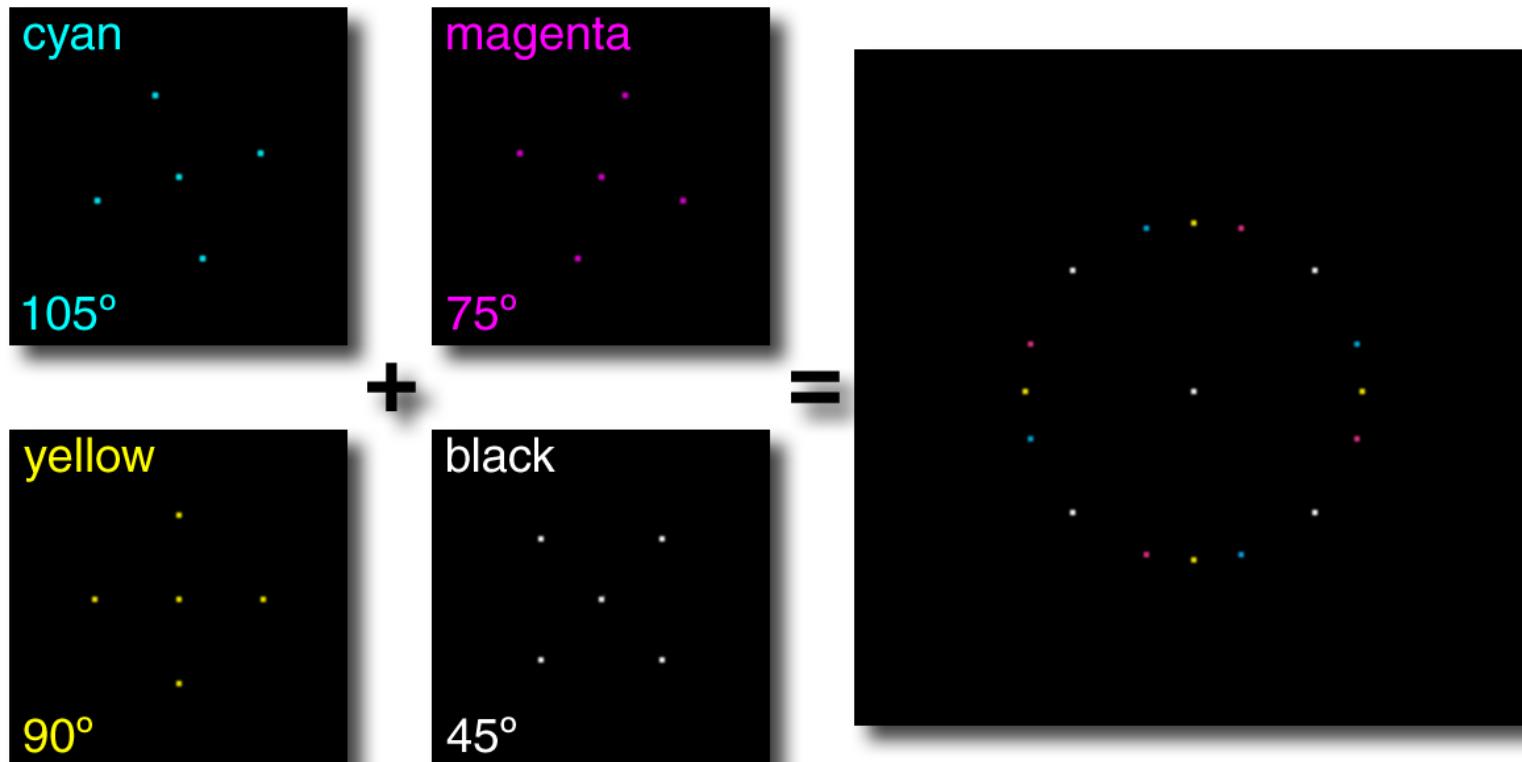
Cyan:	105°
Yellow:	90°
Magenta:	75°
Black:	45°

Image from Adobe Photoshop CS2 documentation.



Each band has 2 perpendicular
sinusoids + a DC component...

CMYK Standard Halftone Screens



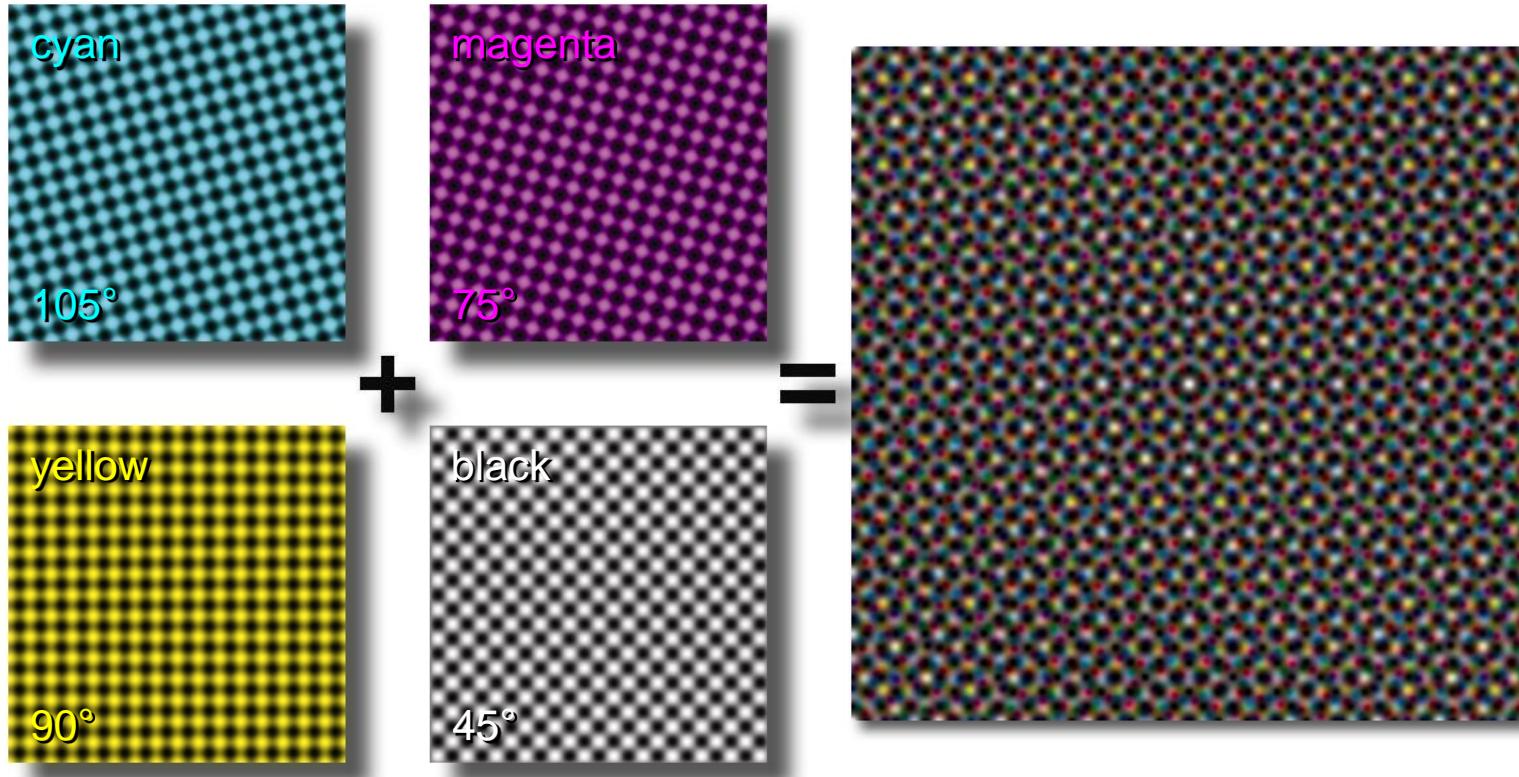
Power Spectra



... which creates rectangular grids at 4 different angles.

EECE 4353 Image Processing
Vanderbilt University School of Engineering

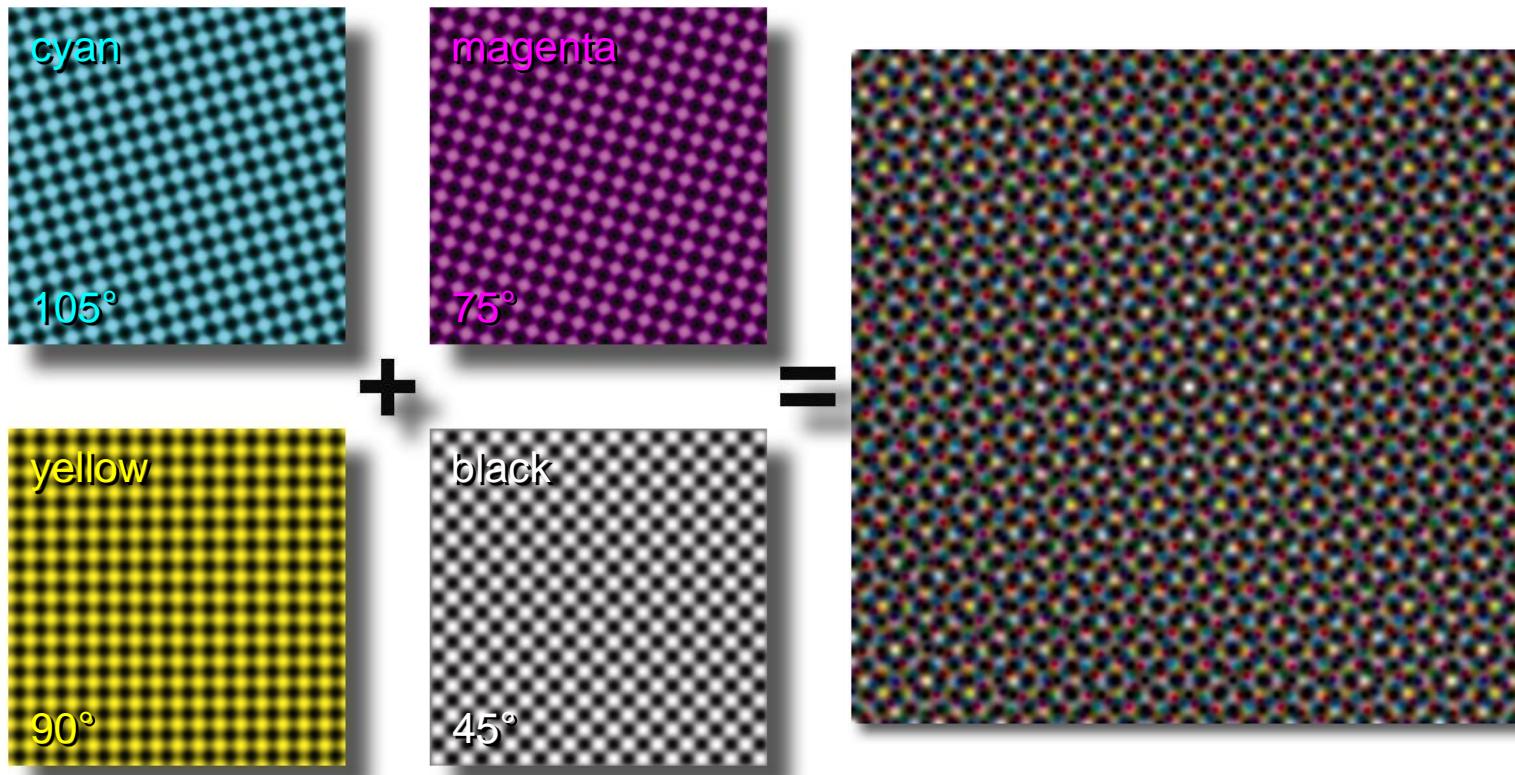
CMYK Standard Halftone Screens





When the 4 are summed, the result is a "rosette" image.

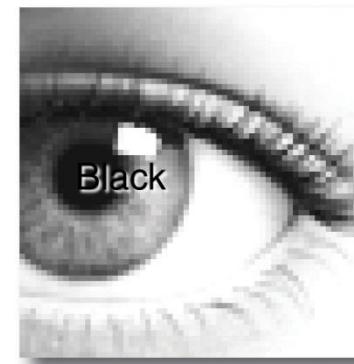
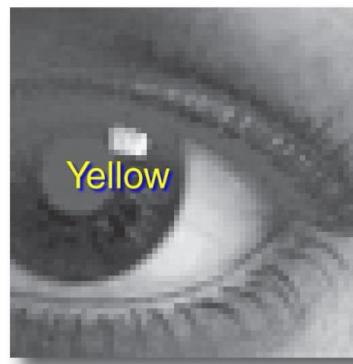
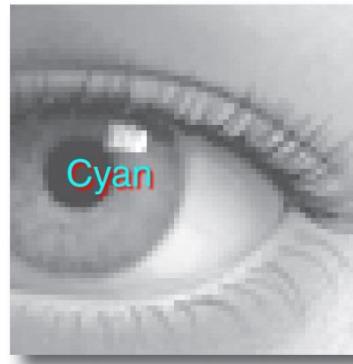
CMYK Standard Halftone Screens





To print an image, it is separated into 4 color bands ...

Example: Color Separation / Halftoning



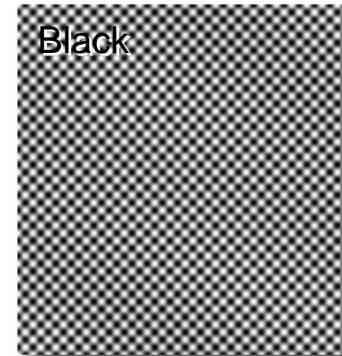
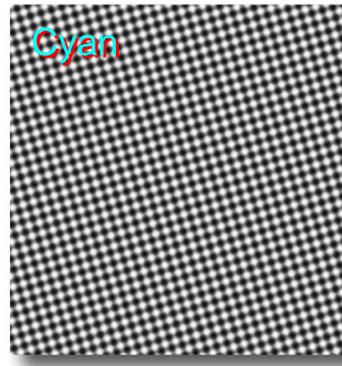
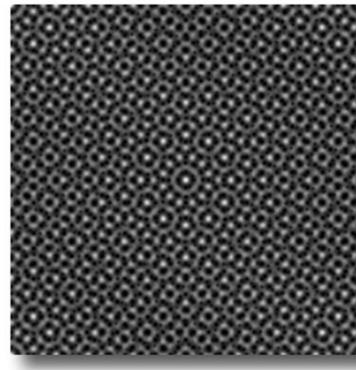
That is, an intensity image is created for each of the four color bands.



... each of which is multiplied by
a corresponding screen.

Color Separation / Halftoning

Each intensity image is multiplied
by a corresponding screen, then

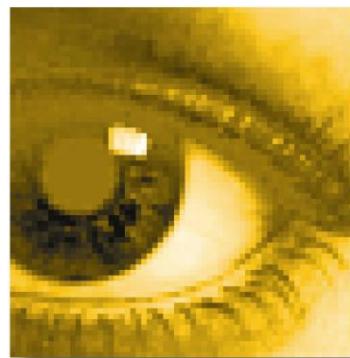
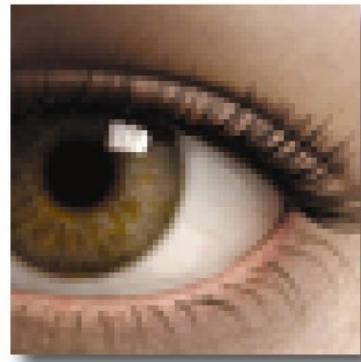


each screened image is printed in
its own color on the same page.



To print an image, it is
separated into 4 color bands ...

Example: Color Separation / Halftoning

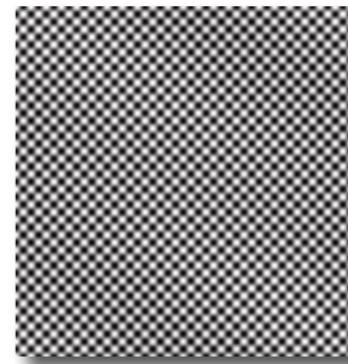
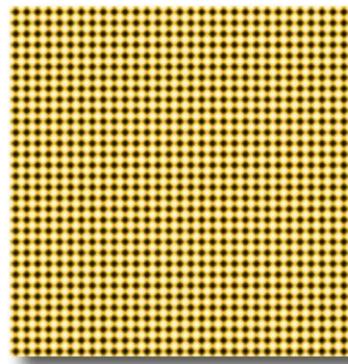
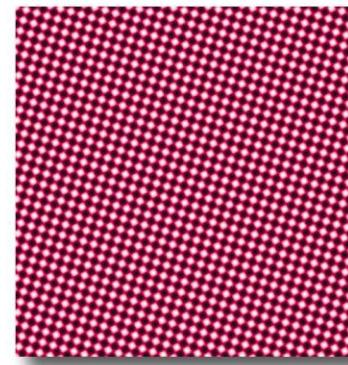
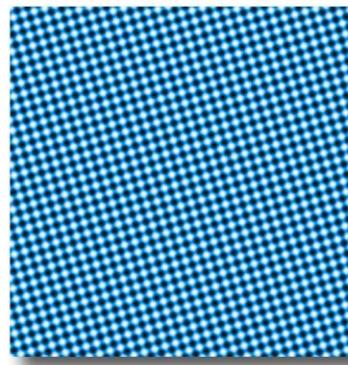
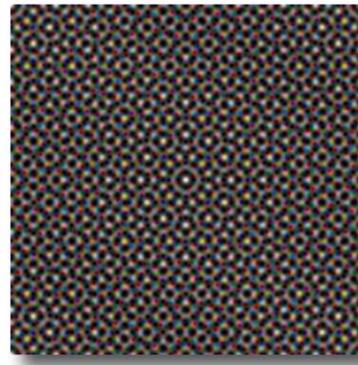


Here the bands tinted in their
corresponding colors.



... each of which is multiplied by
a corresponding screen ...

Example: Color Separation / Halftoning

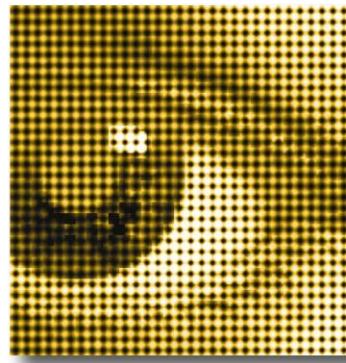
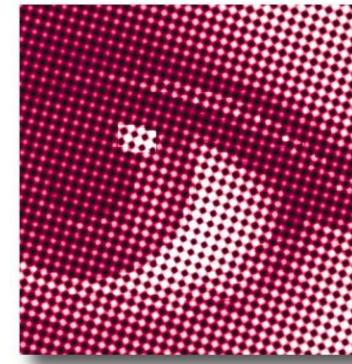
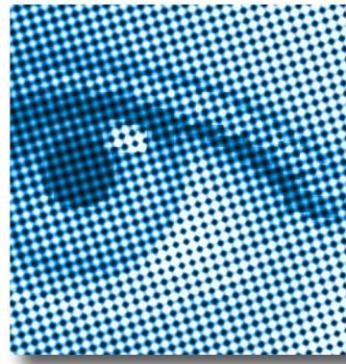
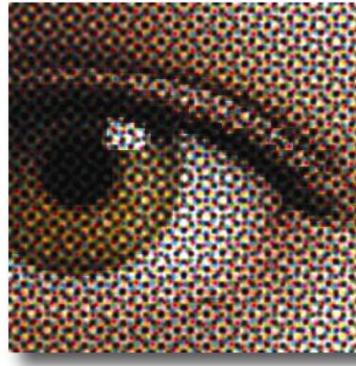


Here the screens tinted in
their corresponding colors.



...to get dot patterns for printing.
The 4 are printed over each other
to get the final result.

Example: Color Separation / Halftoning





Halftone Dots



Image scanned (600 dpi)
from a magazine



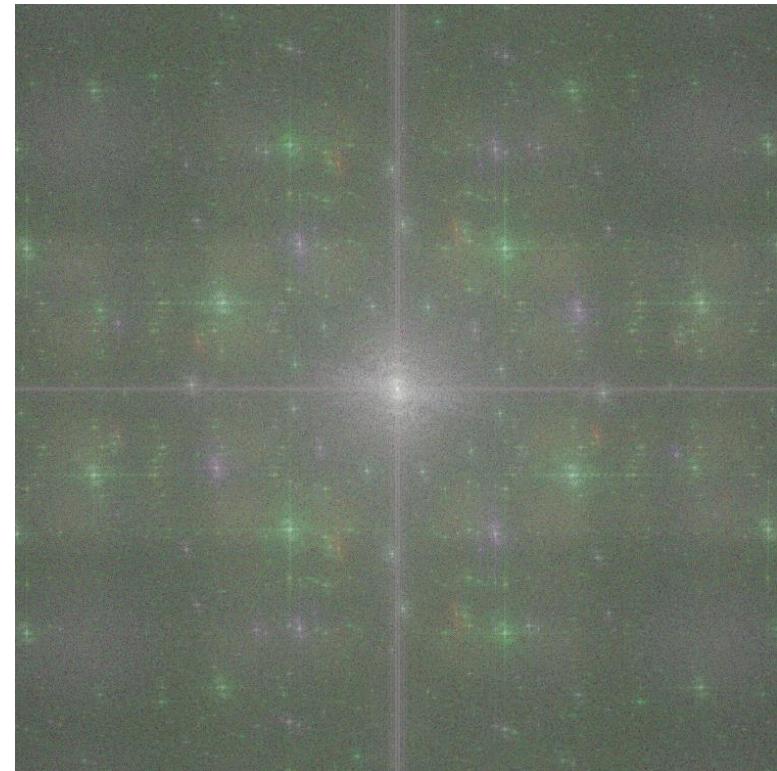
Detail: Circular patterns, the rosettes,
are the result of the halftone dots.



Filtering Out Halftone Dot Distortion



original



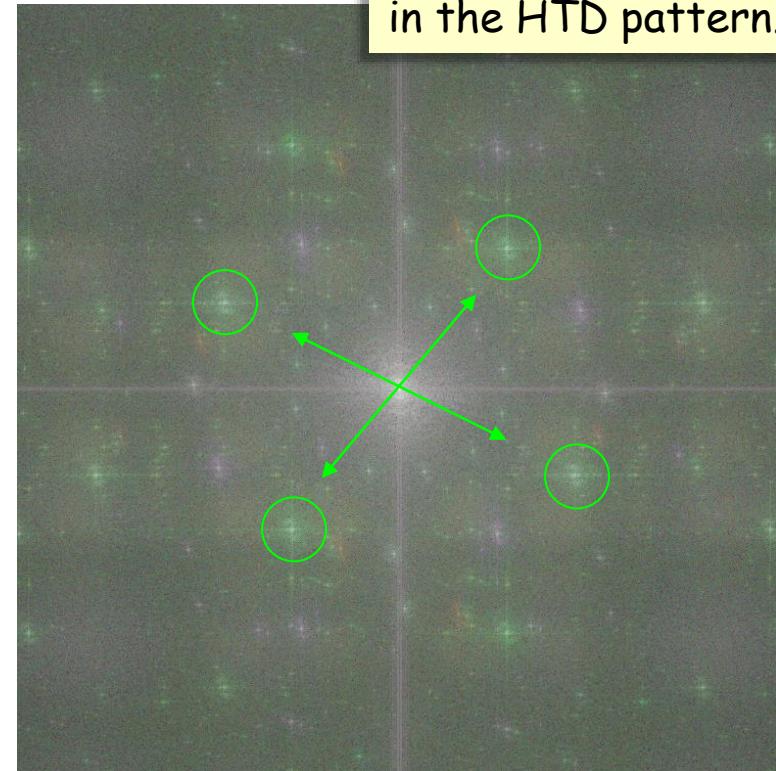
log power spectrum



Filtering Out Halftone Dot D



original



log power spectrum

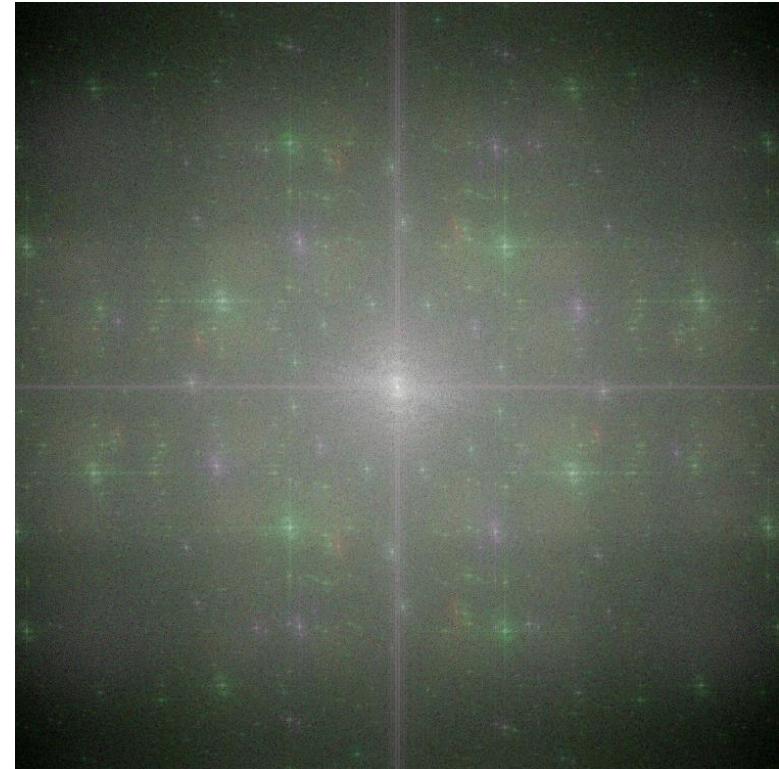
Each pair of peaks corresponds to a sinusoidal sub-pattern in the HTD pattern.



Blurring with a Gaussian ($\sigma = 1$)



blurred image $\sigma=1$



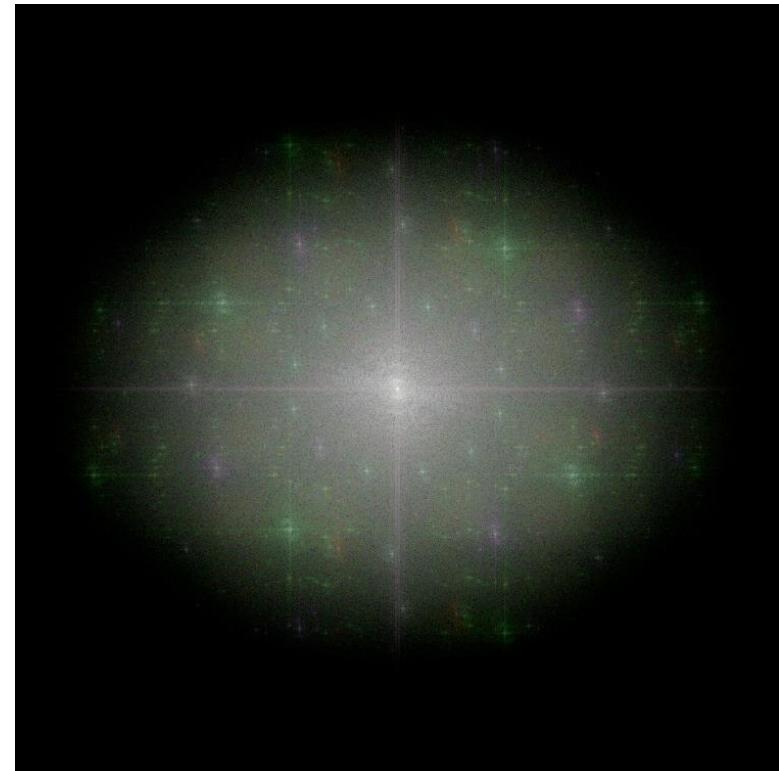
log power spectrum $\sigma=1$



Blurring with a Gaussian ($\sigma = 2$)



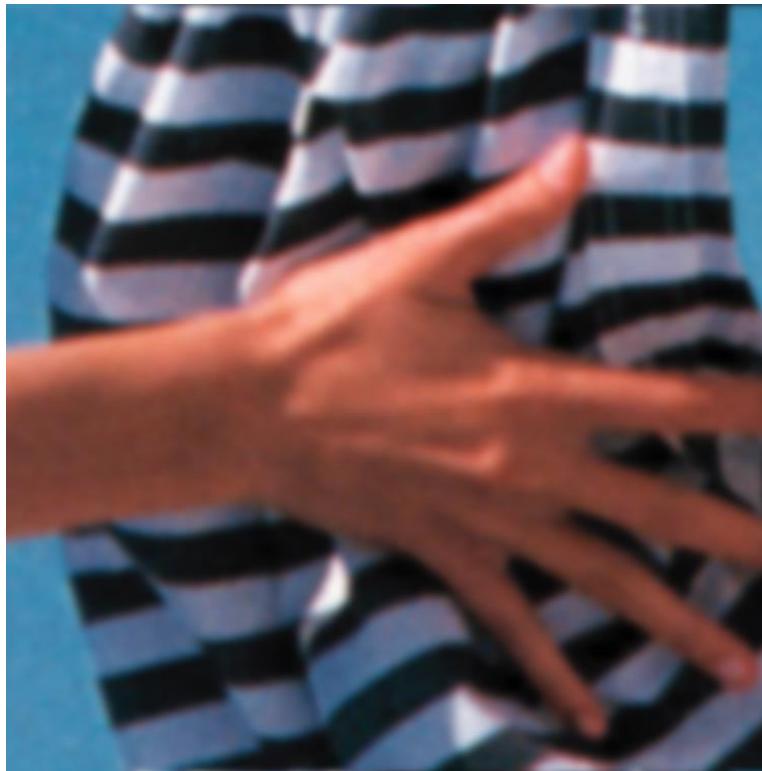
blurred image $\sigma=2$



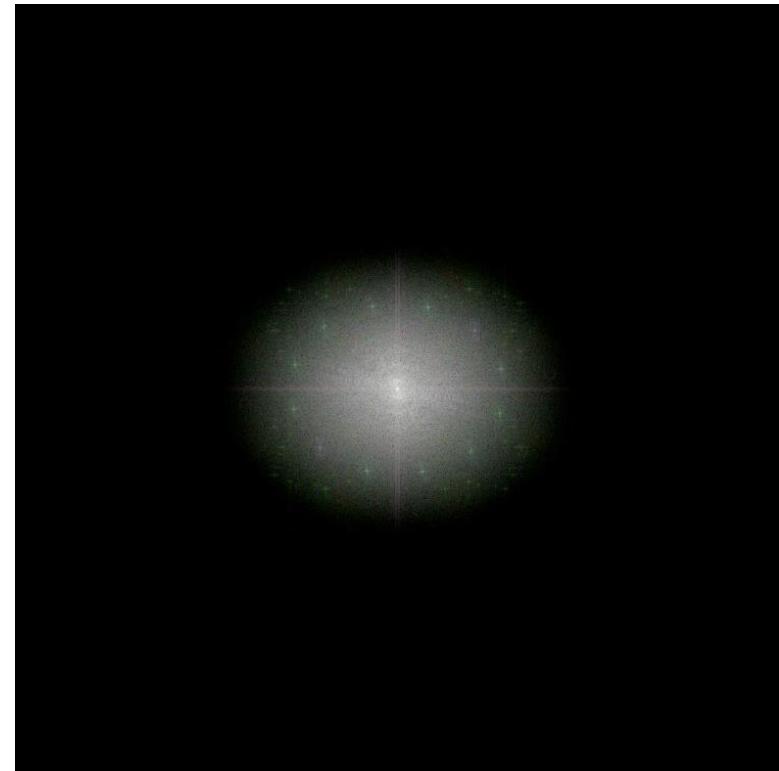
log power spectrum $\sigma=2$



Blurring with a Gaussian ($\sigma = 4$)



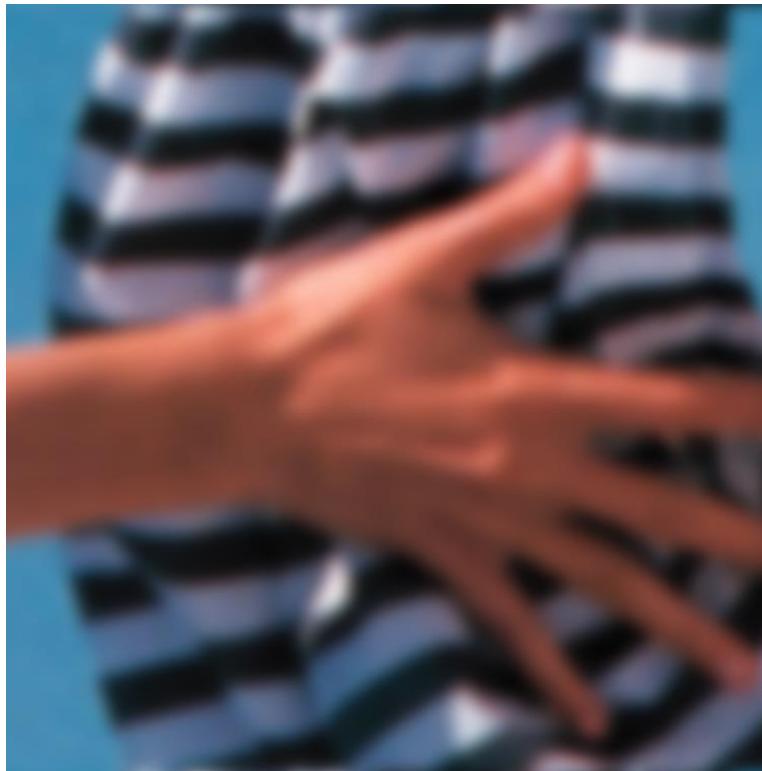
blurred image $\sigma=4$



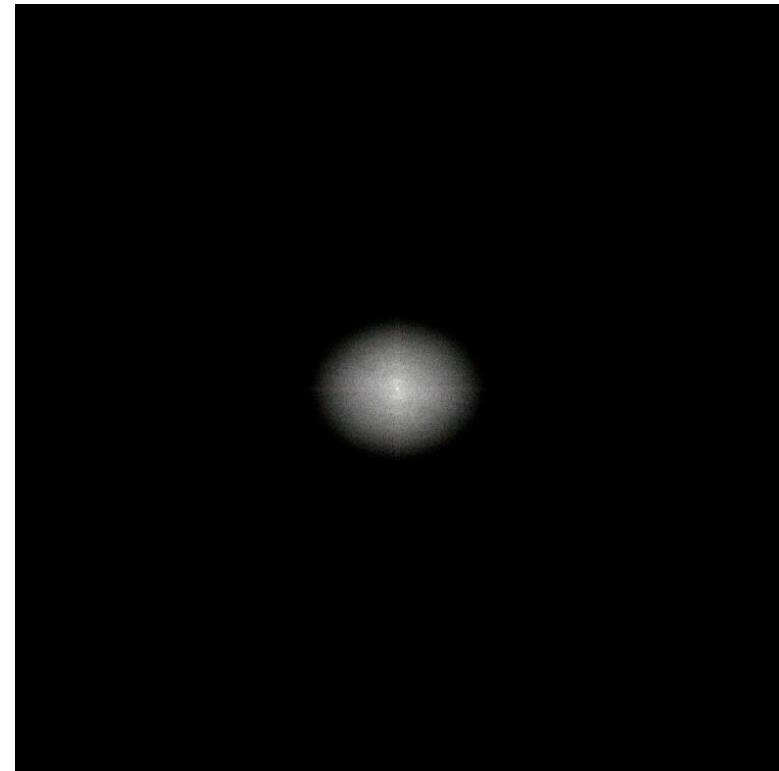
log power spectrum $\sigma=4$



Blurring with a Gaussian ($\sigma = 8$)



blurred image $\sigma=8$



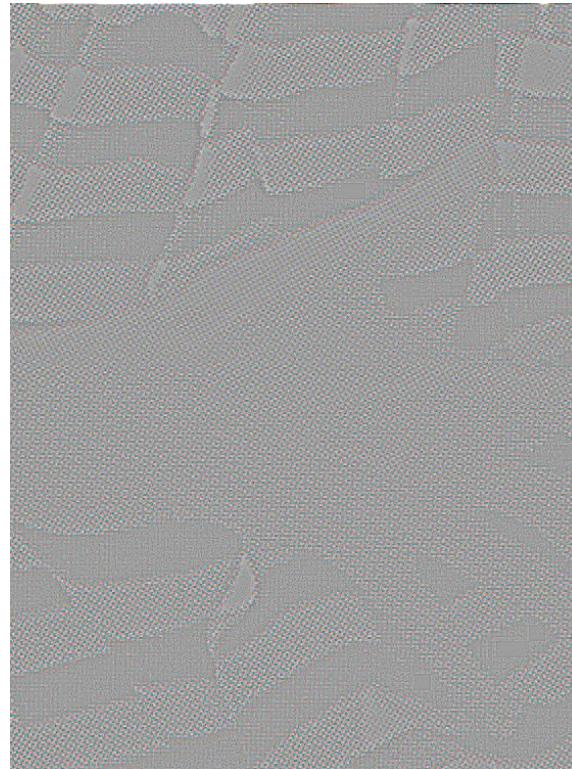
log power spectrum $\sigma=8$



Blurring with a Gaussian ($\sigma = 1$)



original



difference



blurred $\sigma = 1$

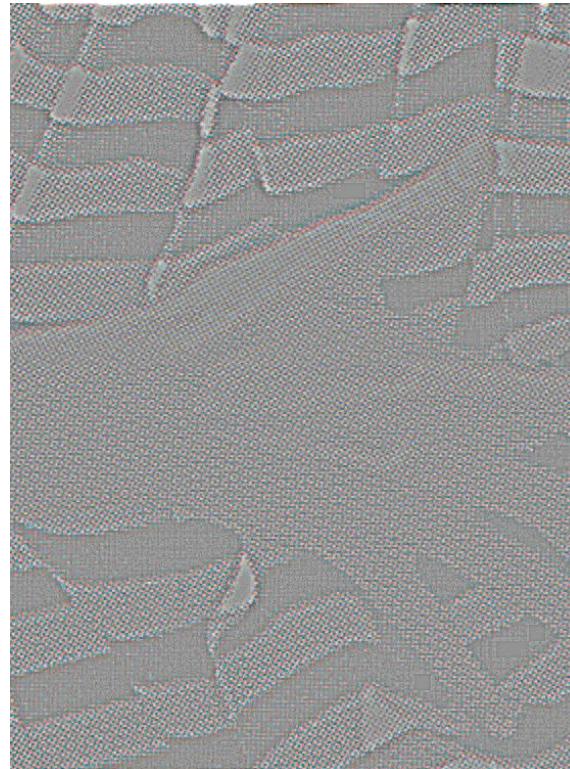
middle gray = 0, normalized



Blurring with a Gaussian ($\sigma = 2$)



blurred $\sigma = 2$



difference

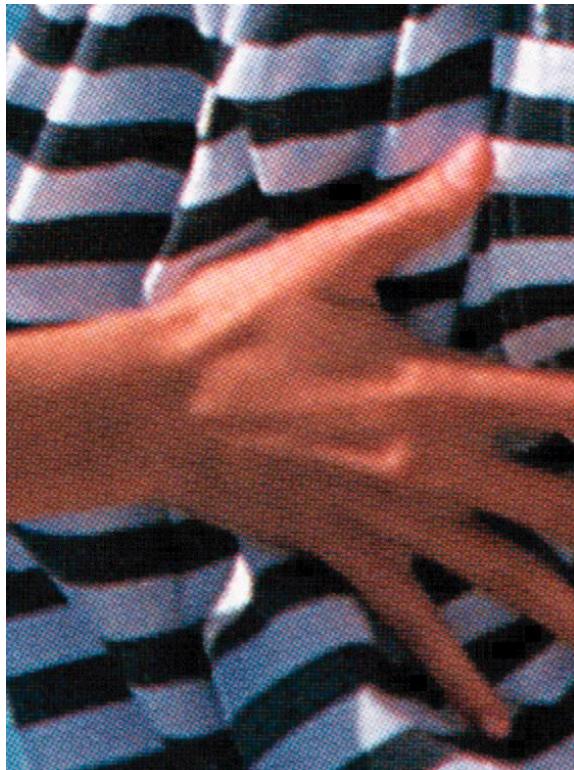
middle gray = 0, normalized



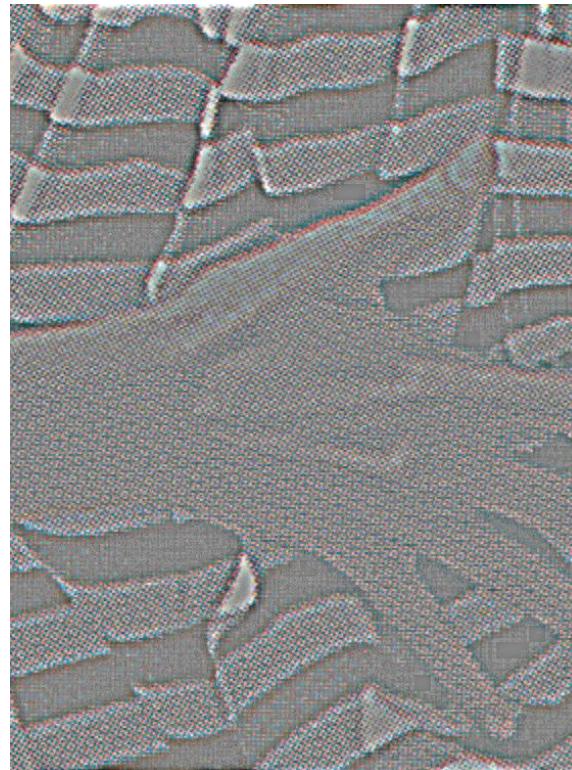
original



Blurring with a Gaussian ($\sigma = 4$)



original



difference



blurred $\sigma = 4$

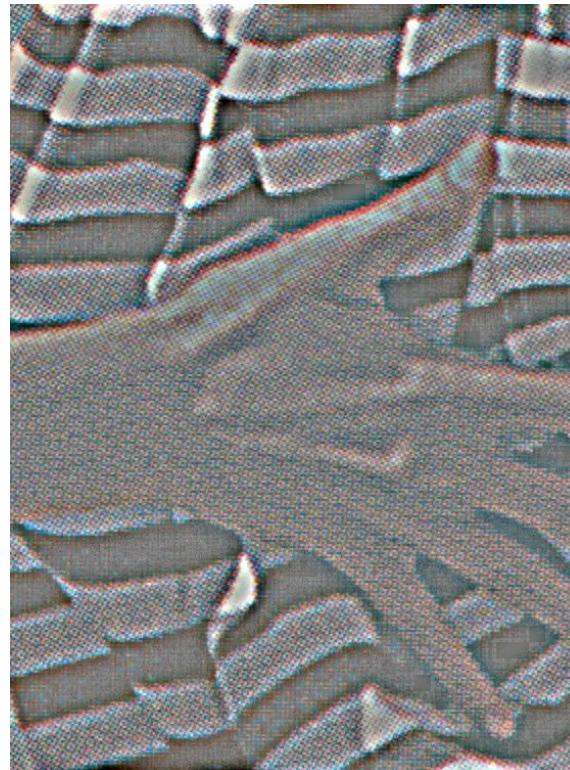
middle gray = 0, normalized



Blurring with a Gaussian ($\sigma = 8$)



blurred $\sigma = 8$



difference

middle gray = 0, normalized



original



Problem with Blurring to Reduce HTD Distortion

⇒ It blurs everything.

Better to remove the HTD frequency components selectively:

1. Read in the image.
2. Compute the log power spectrum of the image.
3. Find the locations of the HTD spectrum peaks.
4. Mark these on a mask.
5. Enlarge the points to regions that cover most of the energy.
6. Blur the mask for used as a notch filter.
7. Multiply the Fourier transform of the image by the mask.
8. Take the inverse Fourier transform of the result.



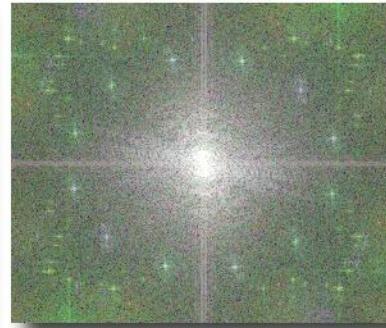
Remove HTD Distortion Selectively

... through notch filtering.

1.



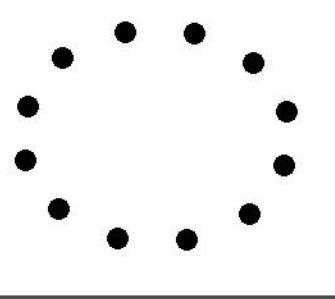
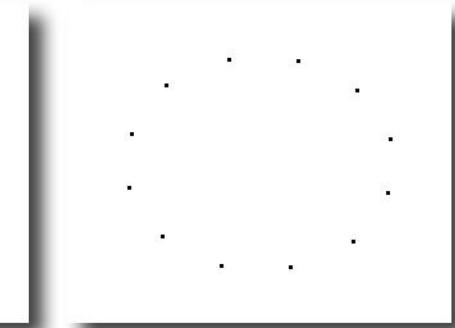
2.



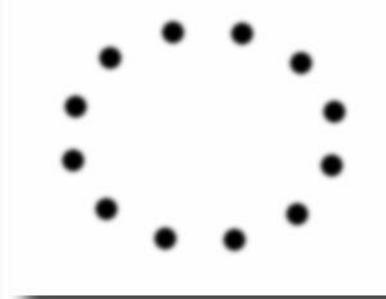
3.

```
HTDlocs = 499 297
          545 320
          571 358
          569 400
          542 438
          493 458
          439 457
          393 434
          367 396
          369 354
          396 316
          445 296
```

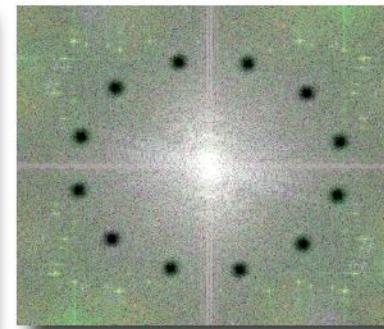
4.



5.



6.



7.



8.

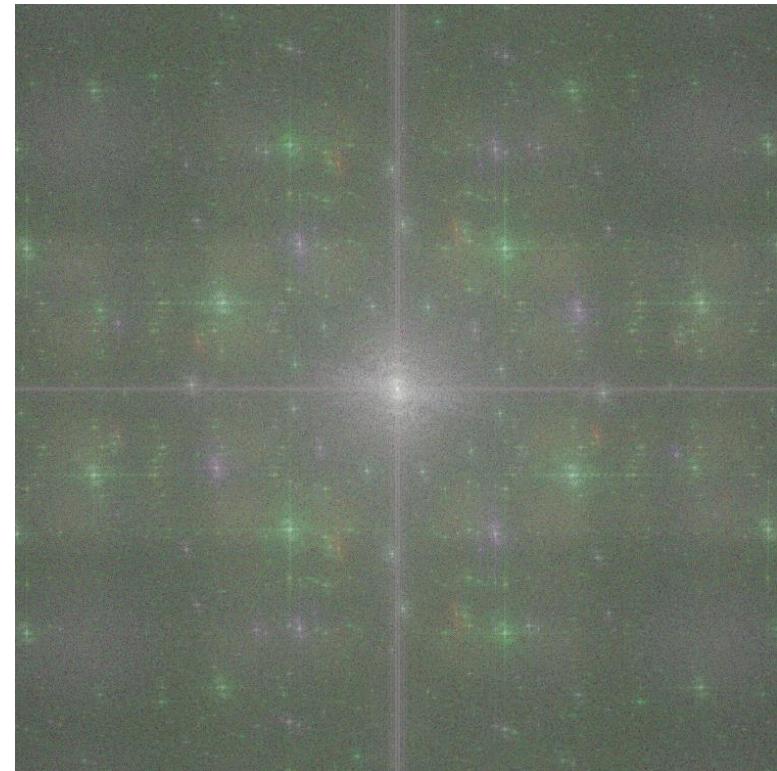
1. Read in image;
2. Compute power spectrum;
3. Locate HTD frequency components;
4. Mark locs on a mask;
5. Enlarge points to regions;
6. Blur the mask;
7. Multiply FT of image by mask;
8. Take inverse FT of result;



Notch Filtering of Halftone Dot Distortion



original



log power spectrum

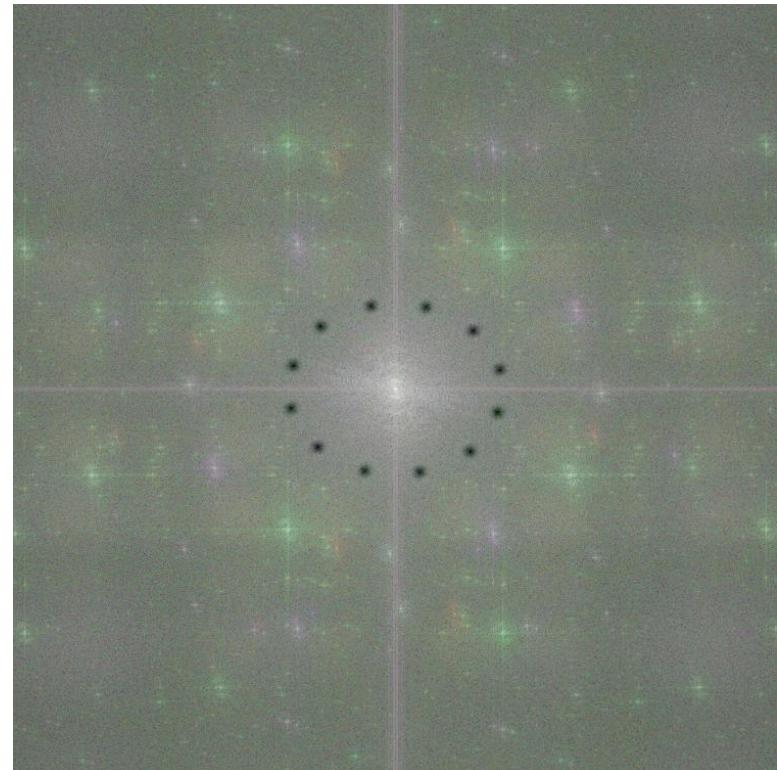


Notch Filtering

Since not much distortion was removed, these must be subharmonics of the true dot frequencies



frequency masked 1



log power spectrum

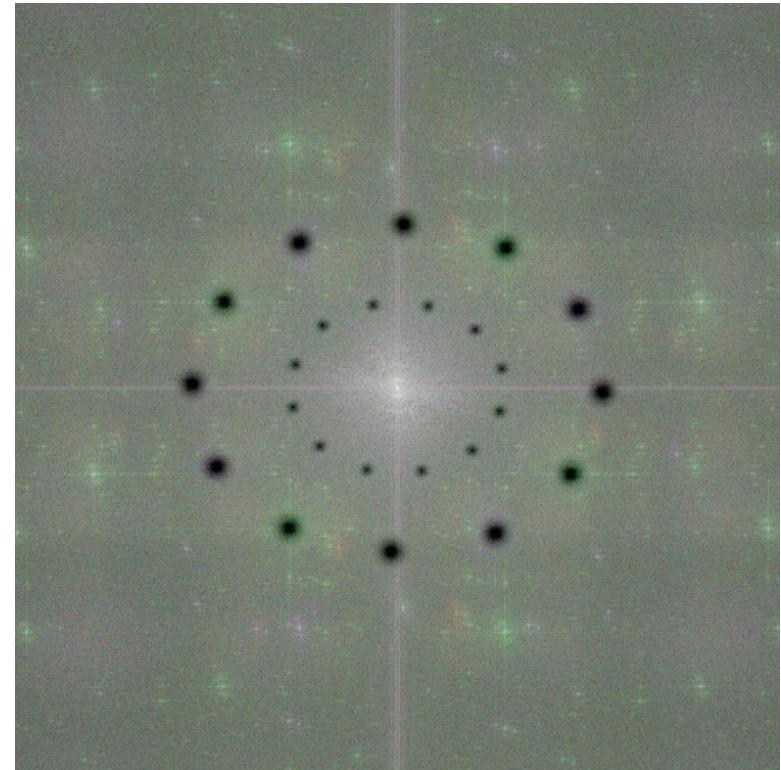


Notch Filtering

The outer ring are the actual HTD frequencies. Can we do any better?



frequency masked 2



log power spectrum

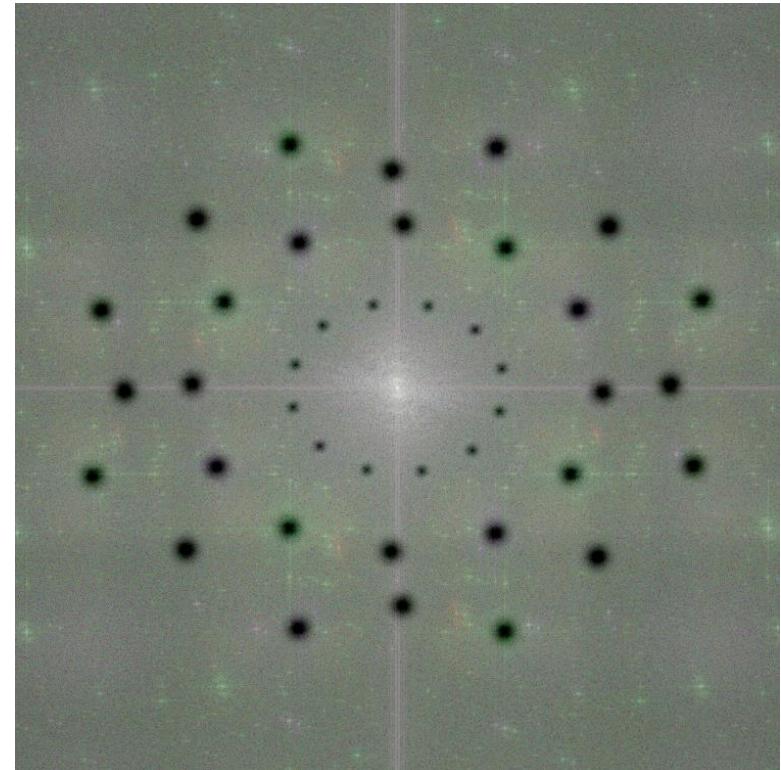


Notch Filtering

Not much. The harmonics contribute little energy to the image.



frequency masked 3



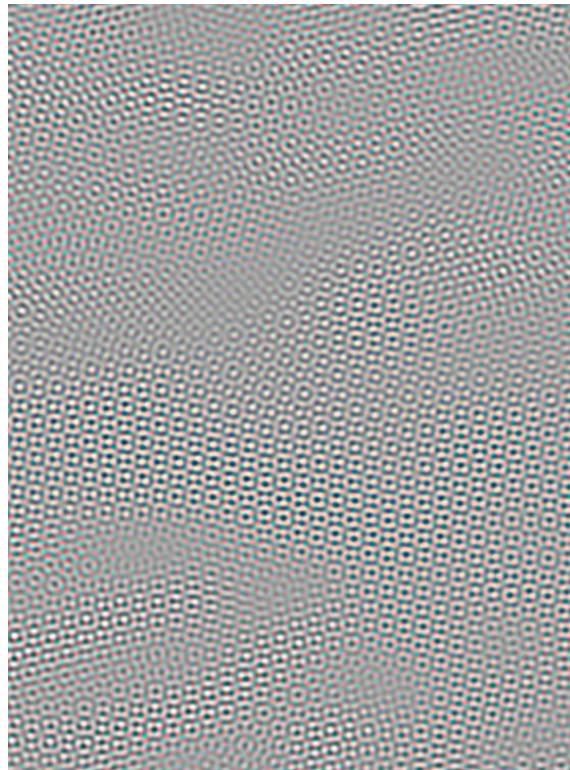
log power spectrum



Notch Filter Difference Images



original



difference



frequency masked 1

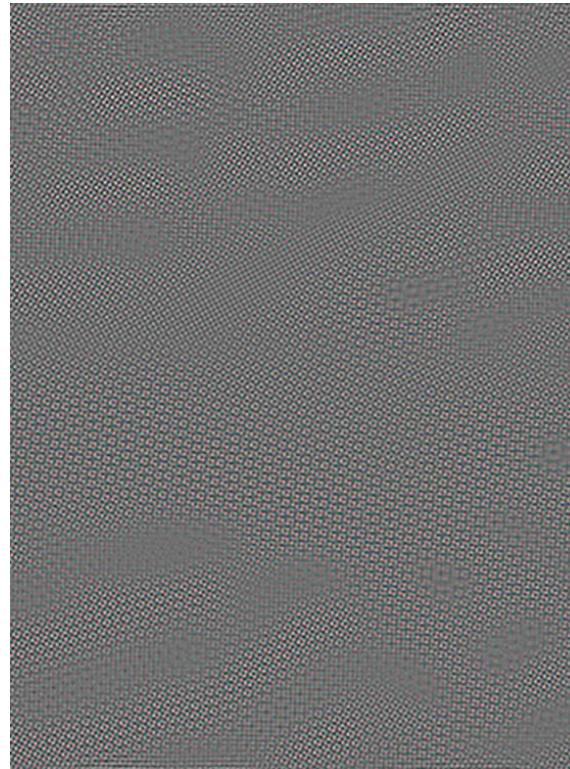
middle gray = 0, normalized



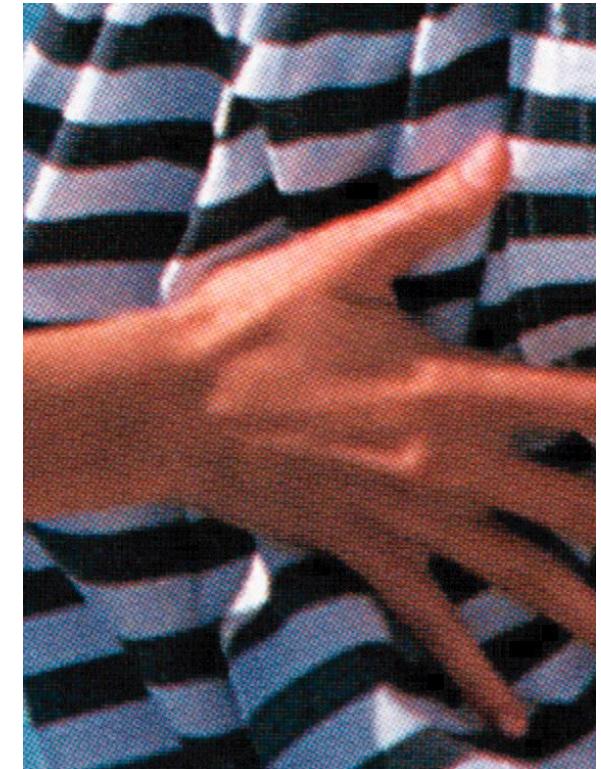
Notch Filter Difference Images



frequency masked 2



difference



original

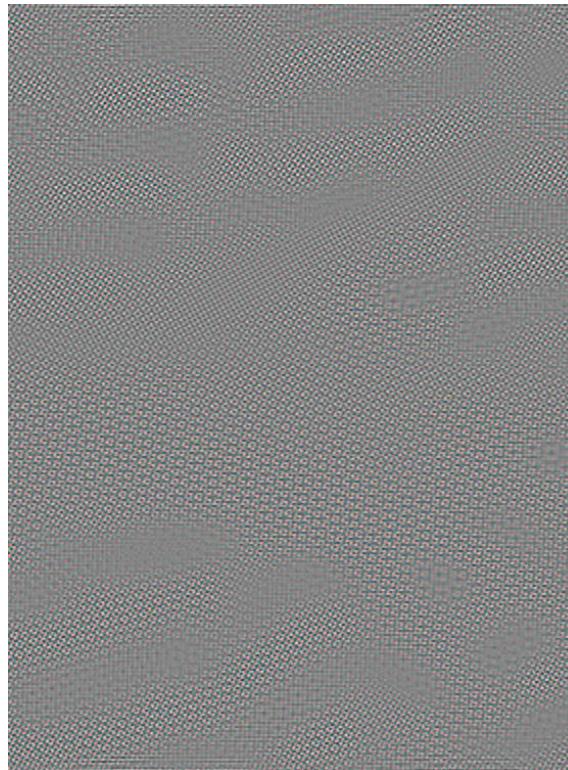
middle gray = 0, normalized



Notch Filter Difference Images



original



difference



frequency masked 3

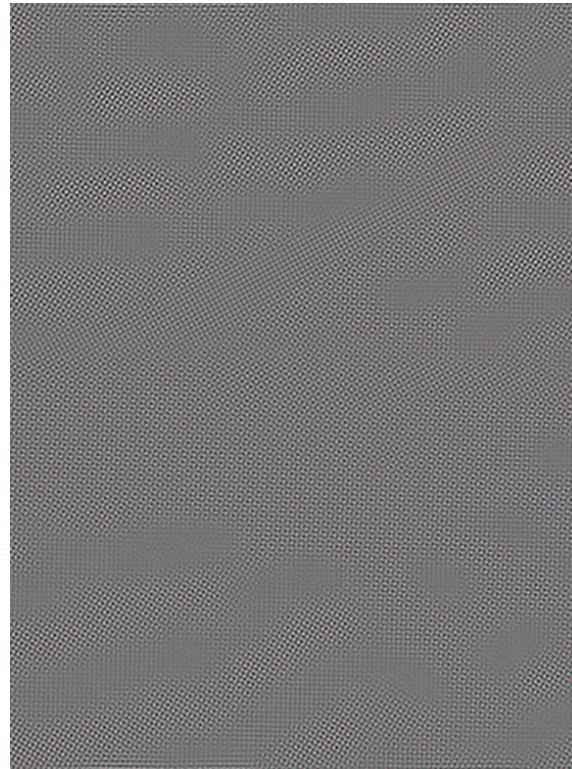
middle gray = 0, normalized



Notch Filter Difference Images



frequency masked 1



difference

middle gray = 0, normalized



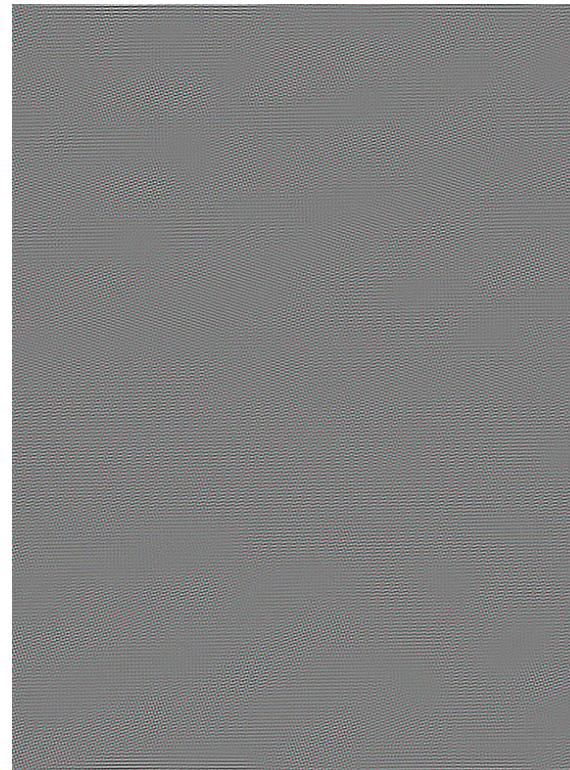
frequency masked 2



Notch Filter Difference Images



frequency masked 2



difference

middle gray = 0, normalized



frequency masked 3