





BLUR, SHARPEN





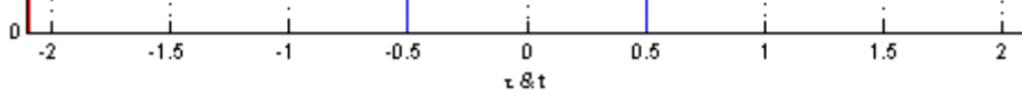
Scene motion

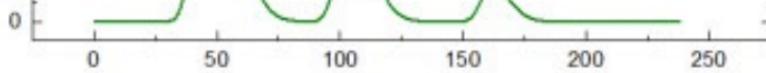
7

BLUR, SHARPEN



$$| * k = B$$





$$f * (g + h) = (f * g) + (f * h)$$

0	1	1	0	0	0	0
1	1	0	0	0	0	0

f (original image)

0	0	1	1	0
0	0	1	1	0

g (kernel)

result

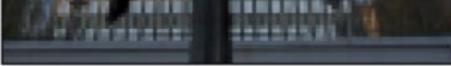
Source: Fredo Durand

(applied in both dimensions)

Fair !!







Horizontal gradient
(absolute value)

Vertical gradient
(absolute value)

Gradient magnitude





You can play with some filters here: <https://setosa.io/ev/image-kernels/>



- But convolutions are expensive... **Fourier domain!**

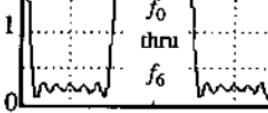


Jean Baptiste Joseph Fourier (1768-1830)





$$f(\text{target}) = f_1 + f_2 + f_3 + \dots + f_n + \dots$$





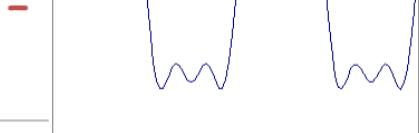


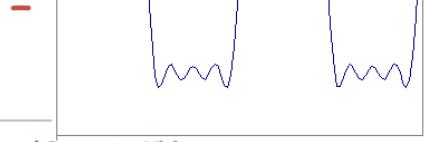
of $f(t)$





Source: James Tompkin









strong-contrast sharp edges!

Source: James Tompkin





and phases (timing) of these sinusoidal components.



Computational Imaging – Master in Robotics, Graphics, and Computer Vision

Amazing slide adapted from James Tompkin's

$$\phi = \tan^{-1} \frac{\text{Im}(\varphi)}{\text{Re}(\varphi)}$$



Further explanations and intuitive interpretation: <https://sites.northwestern.edu/elannesscohn/2019/07/30/developing-an-intuition-for-fourier-transforms/>



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Source: James Tompkin



100 200 300 400 500 600 700 800 900 1000



Fourier transform (amplitude)



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[Source](#)



Fourier transform (amplitude)

Fourier transform (amplitude)



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[Source](#)

Fourier transform (amplitude)



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[Source](#)



[Source](#)

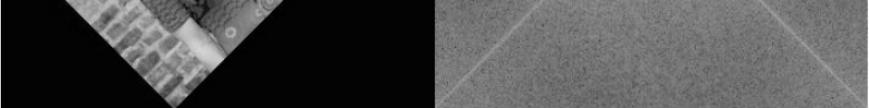


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Source: James Tompkin







[Source](#)



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Source: James Tompkin











the low frequency (smooth) part of the signal can be seen.

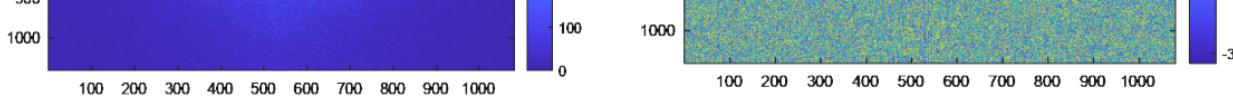


A. Oliva, A. Torralba, P.G. Schyns, "[Hybrid Images](#)," SIGGRAPH 2006



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Source: James Tompkin



Original Image

Reconstruction from only the frequencies' amplitude









- Heavy in low frequencies, talling off in high frequencies



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Source: James Tompkin

Box Filter

Gaussian Filter



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Source: James Tompkin

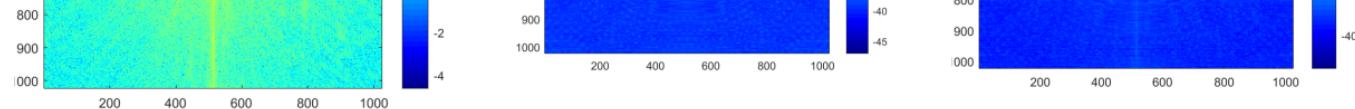
(applied in both dimensions)

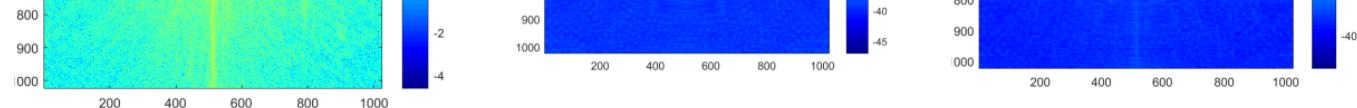
Fair !!

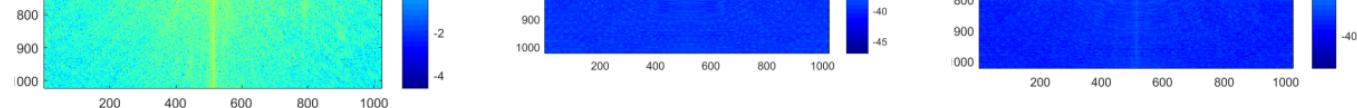


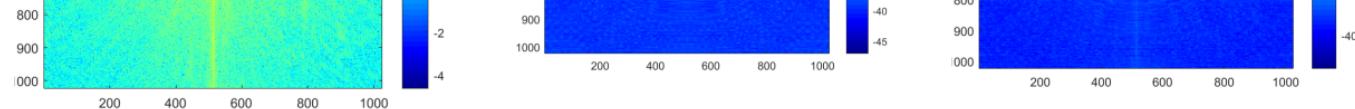


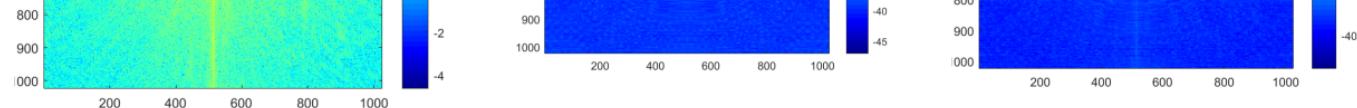












$$B * k^{-1} = |_{est}$$

$$\text{SNR}(\omega) = \frac{\sigma}{\text{noise variance at } \omega}$$

- When SNR is low (high noise), just set to zero.

naïve deconvolution

Wiener deconvolution



• Well... more or less...

