

Administrative

- Delay in Matlab2 grades; Matlab3 comments this week
- Matlab4 due on Thursday
- Trunk schedule change (fix!) for end-of term
 - no exam during finals week (last year's plan)
 - also no quiz 3; usually each quiz is 5% of grade. I'd like to keep each quiz at 5% of grade
- Thursday office hours: 3-4 ok, **can't do 4-5**. When to switch?

Clarifications on Matlab4

- Part 1

- For non-fully-padded case; pad x and h to be same length as the longer of the two
- For the $h(n)$ and $x(n)$ given, you won't be able to hear the time aliasing, but plot of differences should show where it is

- Part 2

- Better to compare DCT and FFT-based compression for 'on-screen' signal
- Note that Matlab's DCT is not just a different normalization, but actually a different variant of the DCT than covered in book. However, Matlab's documentation will give you the formulas you need, as the PDF states

EE-125: Digital Signal Processing

Metrics for DFT-based spectrum analysis

Professor Tracey

Tufts

Reminder: from lecture on windowing

- The DFT/FFT have two main uses
 - Fast FFT-based FIR filtering (overlap/add, etc)
 - Spectrum estimation / spectral analysis
- We may want to do spectral analysis in order to:
 - Learn something about a signal, either by human or automated analysis of the frequency content
 - Do processing in frequency domain (mp3, etc), then go back to time domain
- We'll consider three main topics
 - **Deterministic, non-time-varying signals, possibly in random noise**
 - Time-varying but non-random signals (spectrograms)
 - Random processes / noise (periodograms)

Reminder: Example: 3 sinusoids, 2 closely spaced, Rectangular window

1. When doing DFT, we are applying a window (even if just boxcar)
2. Thus, spectrum estimate is given by convolution:
$$V(\omega) = X(\omega) * W(\omega)$$
3. By increasing the window length, we get smaller mainlobe, and can resolve signals
4. High sidelobes can distort signal & mask weak signals

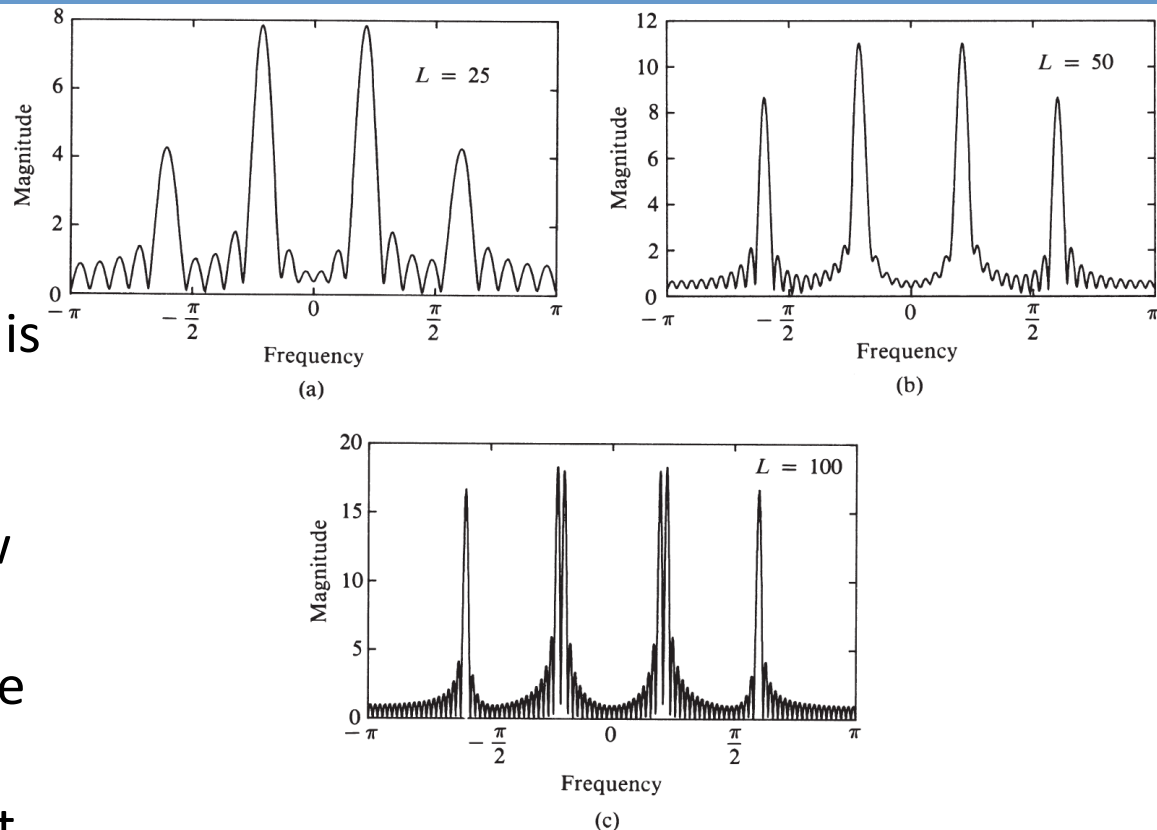


Figure 7.4.2 Magnitude spectrum for the signal given by (7.4.8), as observed through a rectangular window.

Reminder: Example: 3 sinusoids, 2 closely spaced, Hanning window

1. By using other windows, we can suppress sidelobes at cost of widening the main lobe
2. Intuition: window helps make signal look smoothly periodic
3. Mainlobe / sidelobe tradeoff

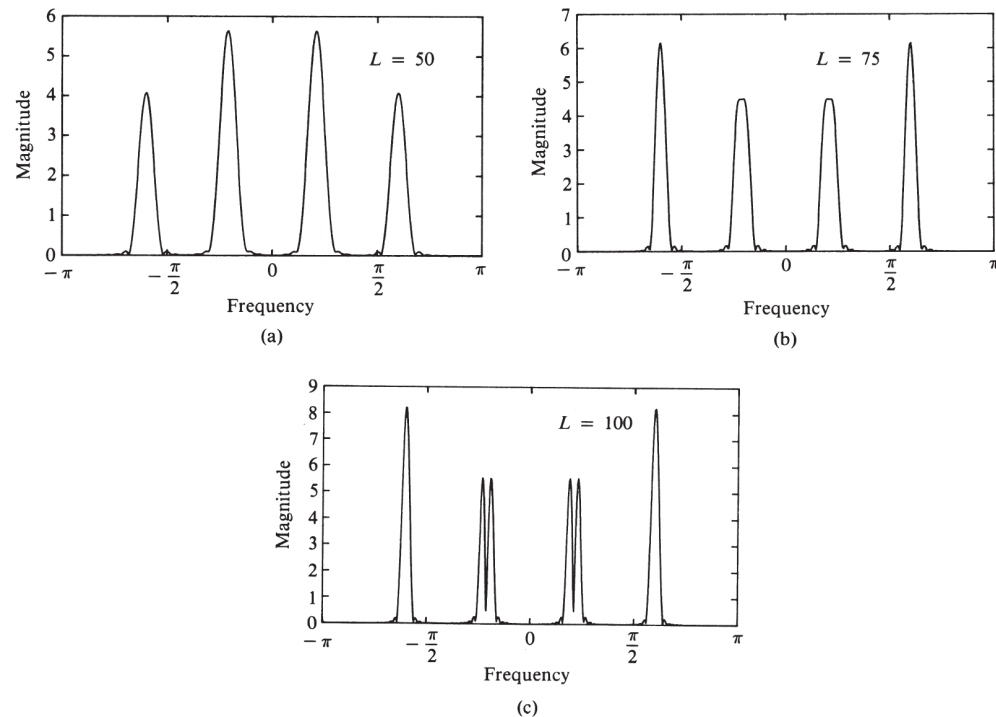


Figure 7.4.4 Magnitude spectrum of the signal in (7.4.8) as observed through a Hanning window.

Some common windows: Table 10.2 in book

Window type	Approx. main lobe width	Peak sidelobe, dB
Boxcar (rectangle)	$4 \pi / M$	-13
Bartlett (triangle)	$8 \pi / M$	-26
Hanning	$8 \pi / M$	-31
Hamming	$8 \pi / M$	-41
Blackman	$12 \pi / M$	-57

Outline for today

- Some common misconceptions
 - Spectrum is really sparse (picket fence effect)
 - Zero-padding improves spectral resolution (i.e., my resolution is what the FFT gives me)
- Harris paper
 - Famous paper on window design for spectrum estimation

Window Figures of Merit:

"On the use of windows for harmonic analysis with the Discrete Fourier Transform," F. Harris, Proc IEEE, 1978.

TABLE I
WINDOWS AND FIGURES OF MERIT

WINDOW	HIGHEST SIDE- LOBE LEVEL (dB)	SIDE- LOBE FALL- OFF (dB/OCT)	COHERENT GAIN	EQUIV NOISE BW (BINS)	3.0-dB BW (BINS)	SCALLOP LOSS (dB)	WORST CASE PROCESS LOSS (dB)	6.0-dB BW (BINS)	OVERLAP CORRELATION (PCNT)	
									75% OL	50% OL
RECTANGLE	-13	-6	1.00	1.00	0.89	3.92	3.92	1.21	75.0	50.0
TRIANGLE	-27	-12	0.50	1.33	1.28	1.82	3.07	1.78	71.9	25.0
$\text{COS}^2(x)$	-23	-12	0.64	1.23	1.20	2.10	3.01	1.65	75.5	31.8
HANNING	-32	-18	0.50	1.50	1.44	1.42	3.18	2.00	65.9	16.7
	-39	-24	0.42	1.73	1.66	1.08	3.47	2.32	56.7	8.5
	-47	-30	0.38	1.94	1.86	0.86	3.75	2.59	48.6	4.3
HAMMING	-43	-6	0.54	1.36	1.30	1.78	3.10	1.81	70.7	23.5
RIESZ	-21	-12	0.67	1.20	1.16	2.22	3.01	1.59	76.5	34.4
RIEMANN	-26	-12	0.59	1.30	1.26	1.89	3.03	1.74	73.4	27.4
DE LA VALLE- POUSSIN	-53	-24	0.38	1.92	1.82	0.90	3.72	2.55	49.3	5.0
TUKEY	-14	-18	0.88	1.10	1.01	2.96	3.39	1.38	74.1	44.4
	-15	-18	0.75	1.22	1.15	2.24	3.11	1.57	72.7	36.4
	-19	-18	0.63	1.36	1.31	1.73	3.07	1.80	70.5	25.1
BOHMAN	-46	-24	0.41	1.79	1.71	1.02	3.54	2.38	54.5	7.4
POISSON	-19	-6	0.44	1.30	1.21	2.09	3.23	1.69	69.9	27.8
	-24	-6	0.32	1.65	1.45	1.46	3.64	2.08	54.8	15.1
	-31	-6	0.25	2.08	1.75	1.03	4.21	2.58	40.4	7.4
HANNING- POISSON	-35	-18	0.43	1.61	1.54	1.26	3.33	2.14	61.3	12.6
	-39	-18	0.38	1.73	1.64	1.11	3.50	2.30	56.0	9.2
	NONE	-18	0.29	2.02	1.87	0.87	3.94	2.65	44.6	4.7
CAUCHY	-31	-6	0.42	1.48	1.34	1.71	3.40	1.90	61.6	20.2
	-35	-6	0.33	1.76	1.50	1.36	3.83	2.20	48.8	13.2
	-30	-6	0.28	2.06	1.68	1.13	4.28	2.53	38.3	9.0
GAUSSIAN	-42	-6	0.51	1.39	1.33	1.69	3.14	1.86	67.7	20.0
	-55	-6	0.43	1.64	1.55	1.25	3.40	2.18	57.5	10.6
	-69	-6	0.37	1.90	1.79	0.94	3.73	2.52	47.2	4.9
DOLPH	-50	0	0.52	1.20	1.22	1.20	3.12	1.85	69.5	23.5

Window Figures of Merit:

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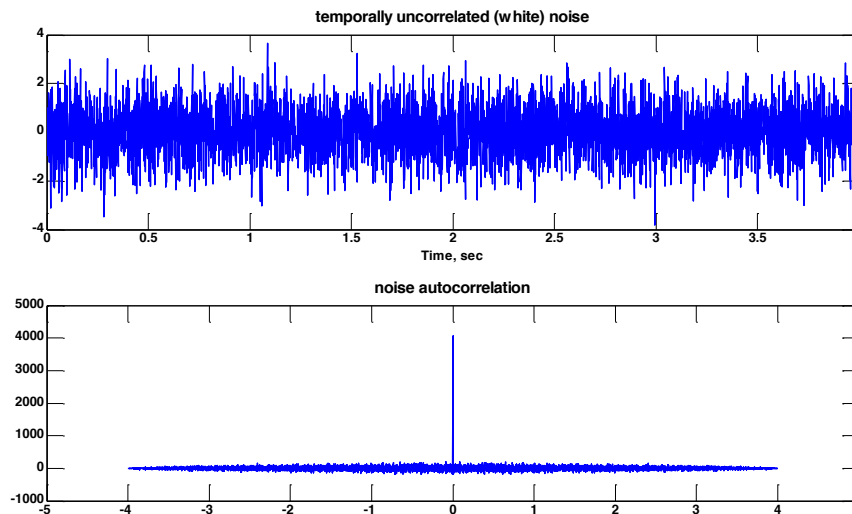
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COS ^a (x) HANNING $a = 1.0$ $a = 2.0$ $a = 3.0$ $a = 4.0$	-23	-12	0.64	1.23	1.20	2.10	3.01	1.65	75.5	31.8
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DOLPH	-50	0	0.52	1.29	1.22	1.20	3.12	1.85	69.5	23.0

Important note; Harris paper assumes no zero-padding

AWGN – Additive Gaussian White Noise

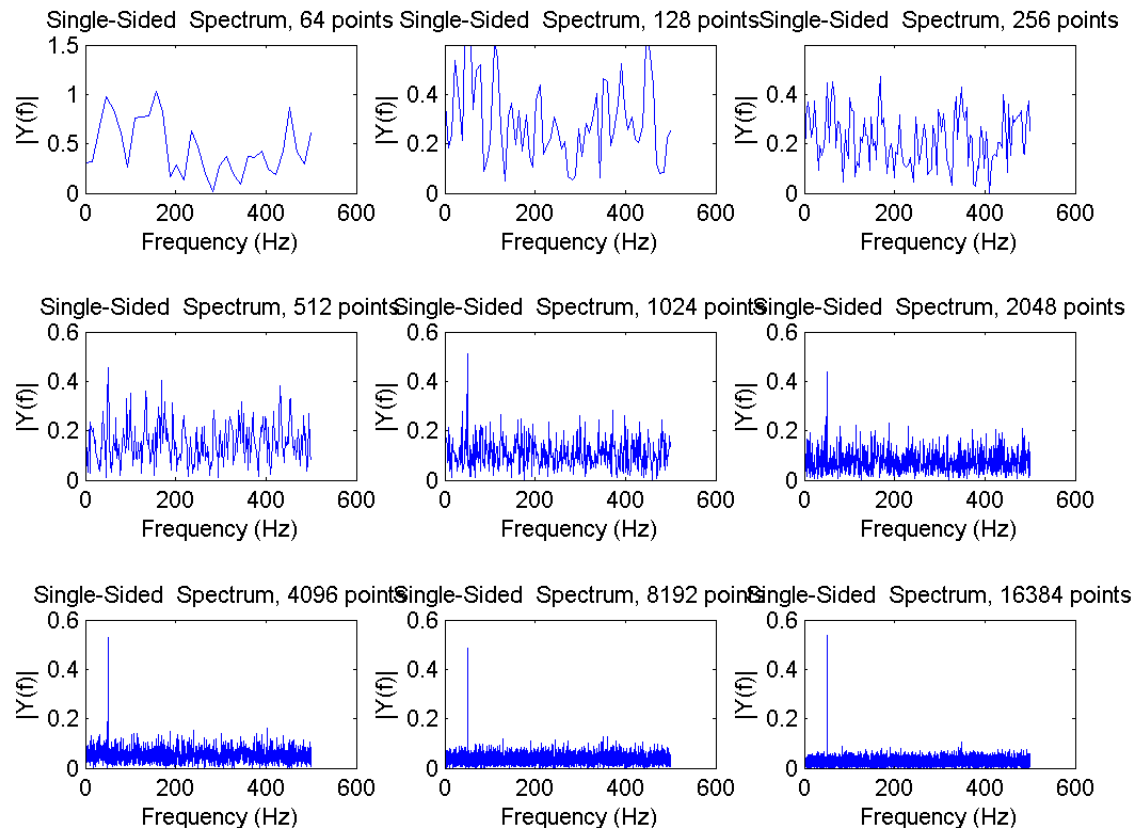
- Additive – added to signal, passes through system
$$y = h*(x+w_{in}) = h*x + w$$
- Gaussian – each individual sample is drawn from a Gaussian distribution: $N(0, \sigma^2)$ (sigma*randn in Matlab)
- White – temporally uncorrelated; each time sample is unrelated to previous or next, so get “white” spectrum



We saw that power spectrum is Fourier transform of autocorrelation

$$\gamma_{ww}(l) = \sigma_w^2 \delta(l)$$

Example of integration time benefit for tonal signal in noise



- Code `integration_time_example.m` uploaded to Trunk