EE 505 Lecture 7

Windowing Spectral Performance of Data Converters

- Time Quantization
- Amplitude Quantization

Clock Jitter Statistical Circuit Modeling

MatLab comparison: 512 Samples with Standard Sweep



Spectre Results

MatLab Results

MatLab comparison: 512 Samples with Strobe Period Sweep



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Considerations for Spectral Characterization

- Tool Validation
- FFT Length
- Importance of Satisfying Hypothesis
 - NP is an integer
 - Band-limited excitation
- Windowing

Windowing - a strategy to address the problem of requiring precisely an integral number of periods to use the DFT for Spectral analysis?

- Windowing is sometimes used
- Windowing is sometimes misused



Example

WLOG assume f_{SIG} =50Hz

 $V_{IN} = sin(\omega t) + 0.5 sin(2\omega t)$ $\omega = 2\pi f_{SIG}$

Consider N_P=20.01 N=4096

Deviation from hypothesis is .05% of the sampling window



Recall

Spectral Response with Non-Coherent Sampling



(zoomed in around fundamental)



Even with N_P =20.001 had significant degradation

Extremely small discontinuity associated with non-coherent sampling causes Significant degradations in spectral response if DFT (and Theorem) used

Windowing

Windowing is the weighting of the time domain function to maintain continuity at the end points of the sample window

Well-studied window functions:

- Rectangular (also with appended zeros)
- Triangular
- Hamming
- Hanning
- Blackman

Sometimes termed a boxcar window

Uniform weight

Can append zeros

Without appending zeros equivalent to no window

Assume f_{SIG} =50Hz

- $V_{IN} = \sin(\omega t) + 0.5 \sin(2\omega t)$
 - $\omega = 2\pi f_{SIG}$

Consider $N_P=20.1$ N=512



Spectral Response with Non-coherent sampling



(zoomed in around fundamental)

Columns 1 through 7

-48.8444 -48.7188 -48.3569 -47.7963 -47.0835 -46.2613 -45.3620

Columns 8 through 14

-44.4065 -43.4052 -42.3602 -41.2670 -40.1146 -38.8851 -37.5520 Columns 15 through 21

-36.0756 -34.3940 -32.4043 -29.9158 -26.5087 -20.9064 -0.1352 Columns 22 through 28

-19.3242 -25.9731 -29.8688 -32.7423 -35.1205 -37.2500 -39.2831 Columns 29 through 35

-41.3375 -43.5152 -45.8626 -48.0945 -48.8606 -46.9417 -43.7344

Columns 1 through 7

-48.8444 -48.7188 -48.3569 -47.7963 -47.0835 -46.2613 -45.3620

Columns 8 through 14

-44.4065 -43.4052 -42.3602 -41.2670 -40.1146 -38.8851 -37.5520

Columns 15 through 21

-36.0756 -34.3940 -32.4043 -29.9158 -26.5087 -20.9064 -0.1352

Columns 22 through 28

-19.3242 -25.9731 -29.8688 -32.7423 -35.1205 -37.2500 -39.2831 Columns 29 through 35

-41.3375 -43.5152 -45.8626 -48.0945 -48.8606 -46.9417 -43.7344 Energy spread over several frequency components

Rectangular Window (with appended zeros)



Triangular Window



Triangular Window





Spectral Response with Non-Coherent Sampling and Windowing



(zoomed in around fundamental)

Triangular Window



Triangular Window

Columns 1 through 7

-100.8530 -72.0528 -99.1401 -68.0110 -95.8741 -63.9944 -92.5170

Columns 8 through 14

-60.3216 -88.7000 -56.7717 -85.8679 -52.8256 -82.1689 -48.3134

Columns 15 through 21

-77.0594 -42.4247 -70.3128 -33.7318 -58.8762 -15.7333 (-6.0918)

Columns 22 through 28

-12.2463 -57.0917 -32.5077 -68.9492 -41.3993 -74.6234 -46.8037

Columns 29 through 35

-77.0686 -50.1054 -77.0980 -51.5317 -75.1218 -50.8522 -71.2410

Hamming Window



Hamming Window





Spectral Response with Non-Coherent Sampling and Windowing



(zoomed in around fundamental)

Comparison with Rectangular Window



Hamming Window

Columns 1 through 7

-70.8278 -70.6955 -70.3703 -69.8555 -69.1502 -68.3632 -67.5133

Columns 8 through 14

-66.5945 -65.6321 -64.6276 -63.6635 -62.6204 -61.5590 -60.4199

Columns 15 through 21

-59.3204 -58.3582 -57.8735 -60.2994 -52.6273 -14.4702 -5.4343

Columns 22 through 28

-11.2659 -45.2190 -67.9926 -60.1662 -60.1710 -61.2796 -62.7277

Columns 29 through 35

-64.3642 -66.2048 -68.2460 -70.1835 -71.1529 -70.2800 -68.1145

Hanning Window



Hanning Window





Spectral Response with Non-Coherent Sampling and Windowing



(zoomed in around fundamental)

Comparison with Rectangular Window



Hanning Window

Columns 1 through 7

-107.3123 -106.7939 -105.3421 -101.9488 -98.3043 -96.6522 -93.0343

Columns 8 through 14

-92.4519 -90.4372 -87.7977 -84.9554 -81.8956 -79.3520 -75.8944

Columns 15 through 21

-72.0479 -67.4602 -61.7543 -54.2042 -42.9597 -13.4511 -6.0601

Columns 22 through 28

-10.8267 -40.4480 -53.3906 -61.8561 -68.3601 -73.9966 -79.0757 Columns 29 through 35

-84.4318 -92.7280 -99.4046 -89.0799 -83.4211 -78.5955 -73.9788

Comparison of 4 windows



Comparison of 4 windows



Preliminary Observations about Windows

- Provide separation of spectral components
- Energy can be accumulated around spectral components
- Simple to apply
- Some windows work much better than others

But – windows do not provide dramatic improvement and ...

Comparison of 4 windows when sampling hypothesis are satisfied


Comparison of 4 windows



Preliminary Observations about Windows

- Provide separation of spectral components
- Energy can be accumulated around spectral components
- Simple to apply
- Some windows work much better than others

But – windows do not provide dramatic improvement and can significantly degrade performance if sampling hypothesis are met

Addressing Spectral Analysis Challenges

- Problem Awareness
- Windowing and Filtering
- Post-processing

Post-processing

Method of circumventing the coherent sampling problem

Can also be used for addressing spectral purity problem for test signal generation



- Easily implemented in MATLAB
- Will be considered in the laboratory
- "Removes" fundamental from samples and replaces with coherent fundamental before taking DFT

Post-processing



- Easily implemented in MATLAB
- Will be considered in the laboratory
- "Removes" fundamental from samples and replaces with coherent fundamental before taking DFT
- Removes spectral impurity of input test signal generator when testing data converters

Issues of Concern for Spectral Analysis

An integral number of periods is critical for spectral analysis

Not easy to satisfy this requirement in the laboratory

Windowing can help but can hurt as well

Out of band energy can be reflected back into bands of interest

Characterization of CAD tool environment is essential

Spectral Characterization of high-resolution data converters requires particularly critical consideration to avoid simulations or measurements from masking real performance

Spectral Characterization of Data Converters

- Distortion Analysis
- Time Quantization Effects
 - of DACs
 - of ADCs
- Amplitude Quantization Effects
 - of DACs
 - of ADCs
- Clock Jitter

time and amplitude depicted

Zero-order sample/hold on DAC or zero-order hold on ADC interpreted output

DAC Assume DAC will be used to generate a continuous time signal Assume DAC is driven by a clock of period T_{CLK}

DAC inputs will be a discrete sequence $\bar{X}(t_k) = \langle X_{quant}(t_k) \rangle$

DAC inputs can change only at times t_k

The duration of each DAC input depends upon system

With zero-order S/H, it is assumed that the DAC output remains constant between transaction times $x_{OUT}(t) = x_{quant}(t_k)$ $t_k \le t < t_{k+1}$



time and amplitude depicted Zero-order sample/hold on DAC or zero-order hold on ADC interpreted output DAC x = (t) - x = (t) $t \le t \le t$



Transition points not necessarily uniformly spaced but will assume so in what follows

time and amplitude depicted

Zero-order sample/hold on DAC or zero-order hold on ADC interpreted output

ADC Output is dimensionless sequence

 $\bar{X}(k) = \langle x_{qant}(t_k) \rangle$

Interpreted output can be represented as a stem plot



time and amplitude depicted Zero-order sample/hold on DAC or zero-order hold on ADC interpreted output



time and amplitude depicted

For zero-order sample/hold on DAC or zero-order hold on ADC interpreted output



(time and amplitude depicted)



(time and amplitude depicted)

16,384 pts res = 4bits



Is this signal band limited?

(time and amplitude depicted)



Simulation environment:

 $N_P=23$ $f_{SIG}=50Hz$ V_{REF} : -1V, 1V Res: will be varied $N=2^n$ will be varied

Spectral Characterization of Data Converters

- Distortion Analysis
- Time Quantization Effects
 - of DACs
 - of ADCs
- Amplitude Quantization Effects
 - of DACs
 - of ADCs
 - Clock Jitter

(time and amplitude depicted)



For amplitude quantization, what appear to be horizontal steps in the above figure are not the same

(amplitude quantization not depicted)











Res = 4 bits



Expect quantization noise effects to be uniformly distributed !!

Res = 4 bits



Expect quantization noise effects to be uniformly distributed !!



Expect quantization noise effects to be uniformly distributed !!



Note presence of odd-ordered harmonic terms !!



Res (n)	SNR _{corr}	SNR
1	3.86	7.78
2	12.06	13.8
3	19.0	19.82
4	25.44	25.84
5	31.66	31.86
6	37.79	37.88
8	49.90	49.92
10	61.95	61.96

Why are there spectral components present in the quantization noise?

Recall the uncorrelated assumption was good only for about 4 bits or more !

Quantization Effects Res = 10 bits



Quantization noise is much more uniform





Harmonic Components not Visible

Quantization Effects Res = 10 bits



Compared to the previous slide, it appears that the quantization noise has gone down – why does this occur?

Res = 10 bits



Compared to the previous slide, it appears that the quantization noise has gone down even more – why does this occur?

Res = 10 bits



Compared to the previous slides, it appears that the quantization noise has gone down even more – why does this occur?



Quantization Effects Res = 10 bits



Very small third harmonic component but does not extend above other noise terms

Spectral Characterization

- Amplitude Quantization

- Does not introduce substantive spectral components for n large
- Nearly uniformly distributed
- Decreases with increasing N
Spectral Characterization of Data Converters

- Distortion Analysis
- → Time Quantization Effects
 - of DACs
 - of ADCs
 - Amplitude Quantization Effects
 - of DACs
 - of ADCs
 - Clock Jitter

Spectral Characteristics of DACs and ADCs



Sampling Clock



Sampled Input Signal (showing time points where samples taken)



Quantized Sampled Input Signal (with zero-order sample and hold)









Consider the following example

- $-f_{SIG}$ =50 Hz
- $-f_{CL}$ =500 Hz (DAC clock)
- f_{DFTCL}=71.24K Hz (coherent sampling)
- $-n_{DFT}=15$ $N=2^{15}=32,768$
- $-N_{P1}$ =23 (number of signal periods in DFT window)
- $-N_{P}=1$
- n_{res}=8 bits
- $-Xin(t) = .95sin(2\pi f_{SIG}t)$ (-.4455dB)
 - Matlab File: afft_Quantization_DAC_Jan2017.m



Expanded View



Width of this region is f_{CL}

Analogous to the overall DFT window when directly sampled but modestly asymmetric



DFT Simulation from Matlab Expanded View



 $(2^{15}=32768)$

DAC Comparisons with Quantization

Fundamental, second harmonic, and third harmonic



Ν	θ	Nsam	n	A ₁	A ₂	A ₃
32K	1	142.5	8	596	-56.7	-64.5
128K	1	569.9	8	596	-56.7	-64.45

Spectral Characteristics of DAC (amplitude and time quantization)

Consider the following example









Expanded View



Consider the following example







Rec Win N=262144 Np =1 Nsam = 1139.75652 nres = 14 fCL/fsig = 10 fDFT/fsig = 11397.5652



Rec Win N=262144 Np =1 Nsam = 1139.75652 nres = 14 fCL/fsig = 10 fDFT/fsig = 11397.5652

Consider the following example



 $-Xin(t) = .95sin(2\pi f_{SIG}t)$ (-.4455dB)





Rec Win N=262144 Np =1 Nsam = 1139.75652 nres = 16 fCL/fsig = 10 fDFT/fsig = 11397.5652



Consider the following example

- $-f_{CL}$ =497.8 Hz (DAC clock)
- $f_{DFTCL} = 141.853K Hz \quad (not coherent sampling)$
- $-n_{DFT}=16$
- $-N_{P1}$ =23 (number of signal periods in DFT window)
- $-N_P=1$
- n_{res}=16bits
- $-Xin(t) = .95sin(2\pi f_{SIG}t)$ (-.4455dB)







Summary of time and amplitude quantization assessment

Time and amplitude quantization do not introduce <u>harmonic</u> distortion

Time and amplitude quantization do increase the noise floor



Stay Safe and Stay Healthy !

End of Lecture 7