

ESE 531: Digital Signal Processing

Lec 1: January 16, 2020
Introduction and Overview



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Lecture Outline

- Course Topics Overview
- Learning Objectives
- Course Structure
- Course Policies
- Course Content
- What is DSP?
- DSP Examples

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Course Topics Overview

- Discrete-Time (DT) Signals
- Time-Domain Analysis of DT Systems
- Discrete Fourier Transform (DFT)
- Fast Fourier Transform (FFT)
- Discrete-Time Fourier Transform (DTFT)
- z-Transform
- Sampling of Continuous Time Signals
- Data Converters and Modulation
- Upsampling/Downsampling
- Discrete-Time Filter Design
- Special Topics

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Learning Objectives

- Learn the fundamentals of digital signal processing
- Provide an understanding of discrete-time signals and systems and digital filters
- Enable you to apply DSP concepts to a wide range of fields
- Gain the ability to read the technical literature on DSP
- Apply the techniques learned in a final project encompassing many different application types

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Learning Objectives

- In other words...

□ Math, Math, Math*

*With MATLAB application for intuition

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Course Structure

- TR Lecture, 4:30-6:00pm in DRLB A2
 - Start 5 minutes after, end 5 minutes early (~75-80min)
- Website (<http://www.seas.upenn.edu/~ese531/>)
 - Course calendar is used for all handouts (lectures slides, assignments, and readings)
 - Canvas used for assignment submission and grades
 - Piazza used for announcements and discussions

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Course Structure

- Course Staff (complete info on course website)
- Instructor: Tania Khanna
 - Office hours – Wednesday 1-3 pm or by appointment
 - Email: taniak@seas.upenn.edu
 - Best way to reach me
- TAs:
 - Dhaval Bhatt
 - Office hours – Th 5-6pm, F 3-5pm
 - Yinghao Zhang
 - Office hours – W 5-6pm, Sat 10am-12pm

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Course Structure

- Lectures
 - Statistically speaking, you will do better if you come to lecture
 - Better if interactive, **everyone** engaged
 - Asking and answering questions
 - Actively thinking about material
- Textbook
 - A. V. Oppenheim and R. W. Schaffer (with J. R. Buck), Discrete-Time Signal Processing, 3rd. Edition, Prentice-Hall, 2010
 - Class will follow text structure... mostly

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Course Structure - Assignments/Exams

- Homework – one week long (~10 total)* [25%]
 - Due Sundays at midnight
 - Combination of book problems and matlab problems
 - Lowest grade dropped
- Project – two weeks long [30%]
 - Work in pairs
 - Combination of different DSP applications
- Midterm exam [20%]
- Final exam [25%]

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Course Policies

See web page for full details

- Turn homework in Canvas
 - Anything handwritten/drawn must be clearly legible
 - Submit code, graphs, test results when specified
 - NO LATE HOMEWORKS!
- Individual work (except project)
 - code, test simulations, analysis, writeups
 - May discuss strategies, but acknowledge help

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Course Content

- Introduction
- Discrete Time Signals & Systems
- Discrete Time Fourier Transform
- Z-Transform
- Inverse Z-Transform
- Sampling of Continuous Time Signals
- Frequency Domain of Discrete Time Series
- Downsampling/Upsampling
- Data Converters, Sigma Delta Modulation
- Frequency Response of LTI Systems
- Signal Flow Representation
- Basic Structures for IIR and FIR Systems
- Design of IIR and FIR Filters
- Butterworth, Chebyshev, and Elliptic Filters
- Filter Banks
- Adaptive Filters
- Computation of the Discrete Fourier Transform
- Fast Fourier Transform

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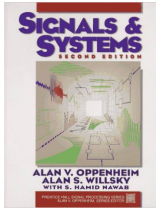
Course Content

Wk	Leet	Date	Lecture	Video	Due	Reading
1	1	1/26 Th	Intro/Overview	Intro/Overview		www course
2	2	1/27 F	Discrete Time Signals & Systems, Part 1			Introduction completely
3	3	1/28 Sa	Discrete Time Signals & Systems, Part 2			2.1-2.2
4	4	1/29 Su	Discrete Time Fourier Transforms			2.3-2.4
5	5	1/30 Mo	Z-Transform			2.5-2.7
6	6	1/31 Tu	Inverse Z-Transform			2.8-3.1
7	7	2/1 F	Sampling and Reconstruction			3.2
8	8	2/2 Tu	Discrete Time Fourier Transforms			3.3-3.4
9	9	2/3 We	Discrete Time Fourier Transforms			3.5-3.7
10	10	2/4 Th	Discrete Time Fourier Transforms			3.8-3.9
11	11	2/5 Fr	Discrete Time Fourier Transforms			3.10-3.11
12	12	2/6 Sa	Discrete Time Fourier Transforms			3.12-3.13
13	13	2/7 Su	Discrete Time Fourier Transforms			3.14-3.15
14	14	2/8 Mo	Discrete Time Fourier Transforms			3.16-3.17
15	15	2/9 Tu	Discrete Time Fourier Transforms			3.18-3.19
16	16	2/10 We	Discrete Time Fourier Transforms			3.20-3.21
17	17	2/11 Th	Discrete Time Fourier Transforms			3.22-3.23
18	18	2/12 Fr	Discrete Time Fourier Transforms			3.24-3.25
19	19	2/13 Sa	Discrete Time Fourier Transforms			3.26-3.27
20	20	2/14 Su	Discrete Time Fourier Transforms			3.28-3.29
21	21	2/15 Mo	Discrete Time Fourier Transforms			3.30-3.31
22	22	2/16 Tu	Discrete Time Fourier Transforms			3.32-3.33
23	23	2/17 We	Discrete Time Fourier Transforms			3.34-3.35
24	24	2/18 Th	Discrete Time Fourier Transforms			3.36-3.37
25	25	2/19 Fr	Discrete Time Fourier Transforms			3.38-3.39
26	26	2/20 Sa	Discrete Time Fourier Transforms			3.40-3.41
27	27	2/21 Su	Discrete Time Fourier Transforms			3.42-3.43
28	28	2/22 Mo	Discrete Time Fourier Transforms			3.44-3.45
29	29	2/23 Tu	Discrete Time Fourier Transforms			3.46-3.47
30	30	2/24 We	Discrete Time Fourier Transforms			3.48-3.49
31	31	2/25 Th	Discrete Time Fourier Transforms			3.50-3.51
32	32	2/26 Fr	Discrete Time Fourier Transforms			3.52-3.53
33	33	2/27 Sa	Discrete Time Fourier Transforms			3.54-3.55
34	34	2/28 Su	Discrete Time Fourier Transforms			3.56-3.57
35	35	2/29 Mo	Discrete Time Fourier Transforms			3.58-3.59
36	36	2/30 Tu	Discrete Time Fourier Transforms			3.60-3.61
37	37	3/1 We	Discrete Time Fourier Transforms			3.62-3.63
38	38	3/2 Th	Discrete Time Fourier Transforms			3.64-3.65
39	39	3/3 Fr	Discrete Time Fourier Transforms			3.66-3.67
40	40	3/4 Sa	Discrete Time Fourier Transforms			3.68-3.69
41	41	3/5 Su	Discrete Time Fourier Transforms			3.70-3.71
42	42	3/6 Mo	Discrete Time Fourier Transforms			3.72-3.73
43	43	3/7 Tu	Discrete Time Fourier Transforms			3.74-3.75
44	44	3/8 We	Discrete Time Fourier Transforms			3.76-3.77
45	45	3/9 Th	Discrete Time Fourier Transforms			3.78-3.79
46	46	3/10 Fr	Discrete Time Fourier Transforms			3.80-3.81
47	47	3/11 Sa	Discrete Time Fourier Transforms			3.82-3.83
48	48	3/12 Su	Discrete Time Fourier Transforms			3.84-3.85
49	49	3/13 Mo	Discrete Time Fourier Transforms			3.86-3.87
50	50	3/14 Tu	Discrete Time Fourier Transforms			3.88-3.89
51	51	3/15 We	Discrete Time Fourier Transforms			3.90-3.91
52	52	3/16 Th	Discrete Time Fourier Transforms			3.92-3.93
53	53	3/17 Fr	Discrete Time Fourier Transforms			3.94-3.95
54	54	3/18 Sa	Discrete Time Fourier Transforms			3.96-3.97
55	55	3/19 Su	Discrete Time Fourier Transforms			3.98-3.99
56	56	3/20 Mo	Discrete Time Fourier Transforms			4.00-4.01
57	57	3/21 Tu	Discrete Time Fourier Transforms			4.02-4.03
58	58	3/22 We	Discrete Time Fourier Transforms			4.04-4.05
59	59	3/23 Th	Discrete Time Fourier Transforms			4.06-4.07
60	60	3/24 Fr	Discrete Time Fourier Transforms			4.08-4.09
61	61	3/25 Sa	Discrete Time Fourier Transforms			4.10-4.11
62	62	3/26 Su	Discrete Time Fourier Transforms			4.12-4.13
63	63	3/27 Mo	Discrete Time Fourier Transforms			4.14-4.15
64	64	3/28 Tu	Discrete Time Fourier Transforms			4.16-4.17
65	65	3/29 We	Discrete Time Fourier Transforms			4.18-4.19
66	66	3/30 Th	Discrete Time Fourier Transforms			4.20-4.21
67	67	3/31 Fr	Discrete Time Fourier Transforms			4.22-4.23
68	68	4/1 Sa	Discrete Time Fourier Transforms			4.24-4.25
69	69	4/2 Su	Discrete Time Fourier Transforms			4.26-4.27
70	70	4/3 Mo	Discrete Time Fourier Transforms			4.28-4.29
71	71	4/4 Tu	Discrete Time Fourier Transforms			4.30-4.31
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77	77	4/10 Mo	Discrete Time Fourier Transforms			4.42-4.43
78	78	4/11 Tu	Discrete Time Fourier Transforms			4.44-4.45
79	79	4/12 We	Discrete Time Fourier Transforms			4.46-4.47
80	80	4/13 Th	Discrete Time Fourier Transforms			4.48-4.49
81	81	4/14 Fr	Discrete Time Fourier Transforms			4.50-4.51
82	82	4/15 Sa	Discrete Time Fourier Transforms			4.52-4.53
83	83	4/16 Su	Discrete Time Fourier Transforms			4.54-4.55
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87	87	4/20 Th	Discrete Time Fourier Transforms			4.62-4.63
88	88	4/21 Fr	Discrete Time Fourier Transforms			4.64-4.65
89	89	4/22 Sa	Discrete Time Fourier Transforms			4.66-4.67
90	90	4/23 Su	Discrete Time Fourier Transforms			4.68-4.69
91	91	4/24 Mo	Discrete Time Fourier Transforms			4.70-4.71
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94	94	4/27 Th	Discrete Time Fourier Transforms			4.76-4.77
95	95	4/28 Fr	Discrete Time Fourier Transforms			4.78-4.79
96	96	4/29 Sa	Discrete Time Fourier Transforms			4.80-4.81
97	97	4/30 Su	Discrete Time Fourier Transforms			4.82-4.83
98	98	5/1 Mo	Discrete Time Fourier Transforms			4.84-4.85
99	99	5/2 Tu	Discrete Time Fourier Transforms			4.86-4.87
100	100	5/3 We	Discrete Time Fourier Transforms			4.88-4.89

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Signals and Systems Review



- ❑ Diagnostic Quiz in Canvas
 - Complete by 1/23 for credit
 - Review in HW 0
 - Also Matlab Tutorial

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What is DSP



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DSP is Everywhere

- ❑ Sound applications
 - Compression, enhancement, special effects, synthesis, recognition, echo cancellation,...
 - Cell phones, MP3 players, movies, dictation, text-to-speech,...
- ❑ Communication
 - Modulation, coding, detection, equalization, echo cancellation,...
 - Cell Phones, dial-up modem, DSL modem, Satellite Receiver,...
- ❑ Automotive
 - ABS, GPS, Active Noise Cancellation, Cruise Control, Parking, Driverless Cars...

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DSP is Everywhere (con't)

- ❑ Medical
 - Magnetic Resonance, Tomography, Electrocardiogram, Biometric Monitoring...
- ❑ Military
 - Radar, Sonar, Space photographs, remote sensing,...
- ❑ Image and Video Applications
 - DVD, JPEG, Movie special effects, video conferencing...
- ❑ Mechanical
 - Motor control, process control, oil and mineral prospecting,...

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Signal Processing

- ❑ Humans are the most advanced signal processors
 - speech and pattern recognition, speech synthesis,...
- ❑ We encounter many types of signals in various applications
 - Electrical signals: voltage, current, magnetic and electric fields,...
 - Mechanical signals: velocity, force, displacement,...
 - Acoustic signals: sound, vibration,...
 - Other signals: pressure, temperature, biometrics...
- ❑ Most real-world signals are analog
 - They are continuous in time and amplitude
 - Convert to voltage or currents using sensors and transducers

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Signal Processing (con't)

- ❑ Analog circuits process these signals using
 - Resistors, Capacitors, Inductors, Amplifiers,...
- ❑ Analog signal processing examples
 - Audio processing in FM radios
 - Video processing in traditional TV sets

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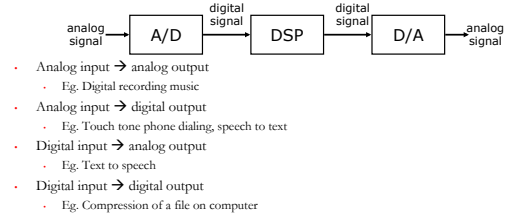
Limitations of Analog Signal Processing

- ❑ Accuracy limitations due to
 - Component tolerances
 - Undesired nonlinearities
- ❑ Limited repeatability due to
 - Tolerances
 - Changes in environmental conditions
 - Temperature
 - Vibration
- ❑ Sensitivity to electrical noise
- ❑ Limited dynamic range for voltage and currents
- ❑ Inflexibility to changes
- ❑ Difficulty of implementing certain operations
 - Nonlinear operations
 - Time-varying operations
- ❑ Difficulty of storing information

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Digital Signal Processing

- ❑ Represent signals by a sequence of numbers
 - Sampling and quantization (or analog-to-digital conversion)
- ❑ Perform processing on these numbers with a digital processor
 - Digital signal processing
- ❑ Reconstruct analog signal from processed numbers
 - Reconstruction or digital-to-analog conversion



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Pros and Cons of Digital Signal Processing

- ❑ Pros
 - Accuracy can be controlled by choosing word length
 - Repeatable
 - Sensitivity to electrical noise is minimal
 - Dynamic range can be controlled using floating point numbers
 - Flexibility can be achieved with software implementations
 - Non-linear and time-varying operations are easier to implement
 - Digital storage is cheap
 - Digital information can be encrypted for security
 - Price/performance and reduced time-to-market

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Pros and Cons of Digital Signal Processing

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- ❑ Cons
 - Sampling causes loss of information
 - A/D and D/A requires mixed-signal hardware
 - Limited speed of processors
 - Quantization and round-off errors

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DSP Examples



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Example I: Audio Compression

- ❑ Compress audio by 10x without perceptual loss of quality
- ❑ Sophisticated processing based on models of human perception
- ❑ 3MB files instead of 30MB
 - Entire industry changed in less than 10 years!

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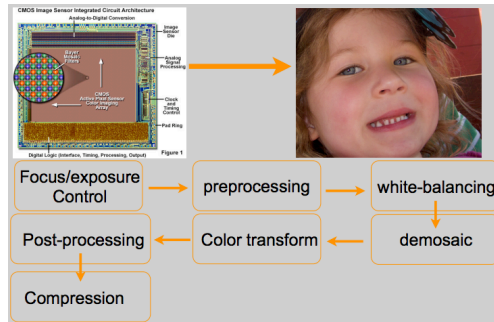
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Historical Forms of Compression

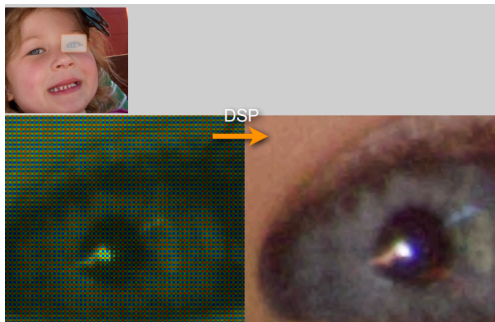
- Morse code: dots (1 unit) dashes (3 units)
 - Code Length inversely proportional to frequency of character
 - E (12.7%) = . (1 unit) Q (0.1%) = --.- (10 units)
- “92 Code”
 - Used by Western-Union in 1859 to reduce BW on telegraph lines by numerical codes for frequently used phrases
 - 1 = wait a minute
 - 73 = Best Regards
 - 88 = Loves and Kisses

73 Best Regards 19units
 73 Best Regards 59units

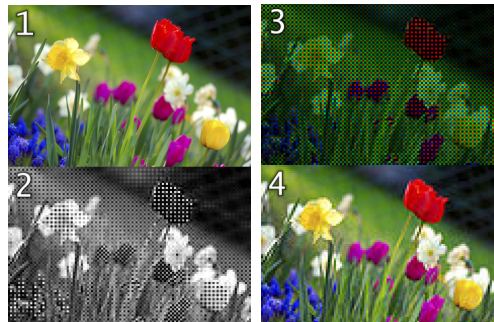
Example II: Digital Imaging Camera



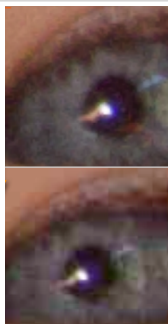
Example II: Digital Imaging Camera



Example II: Digital Imaging Camera

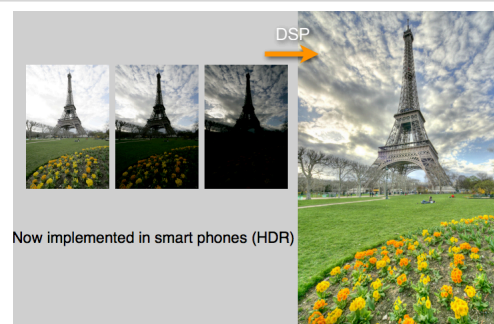


Example II: Digital Imaging Camera



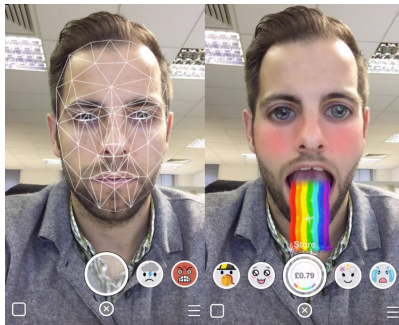
- Compression of 40x without perceptual loss of quality.
- Example of slight over compression: difference enables 60x compression!

Computational Photography



Now implemented in smart phones (HDR)

Image Processing



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Image Processing - Saves Lives

Canadian 'swirl face'

jailed in Thailand



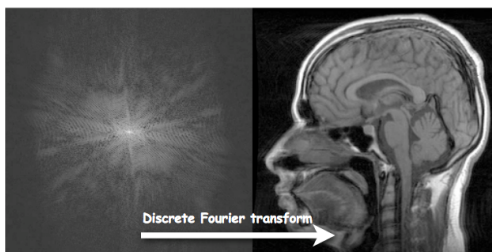
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Example III: MRI

k-space (raw data)

Image

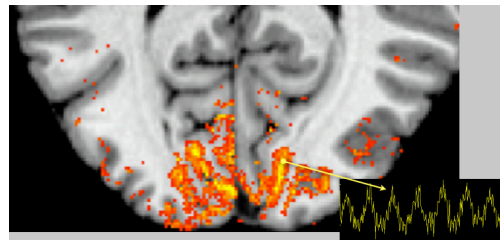


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fMRI example

- Sensitivity to blood oxygenation
 - response to brain activity Convert from one signal to another

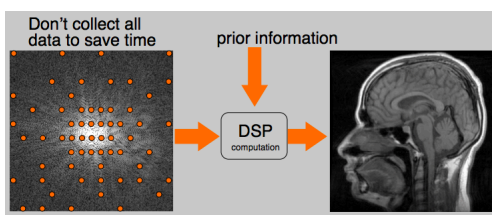


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Compressive Sampling

- Compression meets sampling

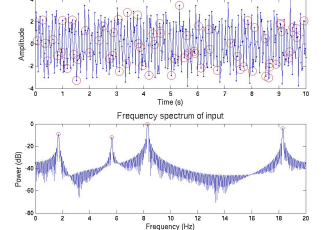


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Example: Sum of Sinusoids

Input signal with undersampled measurements circled (~17.5% of Nyquist samples)



- Sense signal randomly M times
 - $M > C \mu 2(\Phi, \Psi) S \log N$
- Recover with linear program

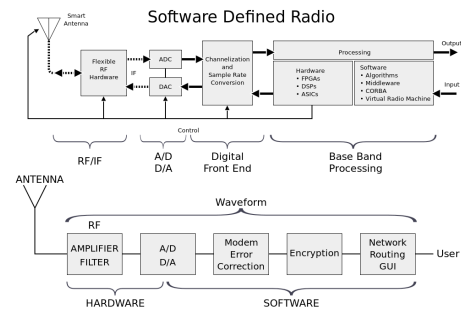
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Example IV: Software Defined Radio

- ❑ Traditional radio:
 - Hardware receiver/mixers/demodulators/filtering
 - Outputs analog signals or digital bits
- ❑ Software Defined Radio:
 - Uses RF front end for baseband signal
 - High speed ADC digitizes samples
 - All processing chain done in software

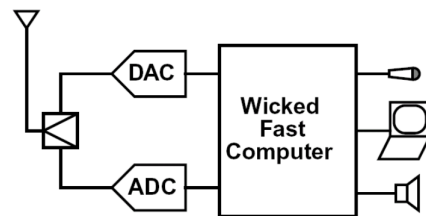
Software Defined Radio



Software Defined Radio

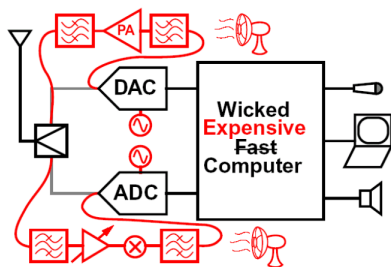
- ❑ Advantages:
 - Flexibility
 - Upgradable
 - Sophisticated processing
 - Ideal Processing chain
 - not approximate like in analog hardware
- ❑ Already used in consumer electronics
 - Cellphone baseband processors
 - Wifi, GPS, etc....

Software Radio Vision



[Schreier, "ADCs and DACs: Marching Towards the Antenna," GIRAFE workshop, ISSCC 2003]

Software Radio Reality



[Schreier, "ADCs and DACs: Marching Towards the Antenna," GIRAFE workshop, ISSCC 2003]

Shameless Plug

- ❑ If you are interested in how Analog to digital converters work and how to make them
- ❑ Take ESE 568!
- ❑ Good to know both sides of the system

Future of ADC design

- Today's ADCs are extremely well optimized
- For non-incremental improvements, we must explore new ideas in signal processing that tackle ADC inefficiency at the system level
 - Compressed sensing
 - Finite innovation rate sampling
 - Other ideas?

Filter Design Example



Optimal Filter Design

- Window method
 - Design Filters heuristically using windowed sinc functions
- Optimal design
 - Design a filter $h[n]$ with $H(e^{j\omega})$
 - Approximate $H_d(e^{j\omega})$ with some optimality criteria - or satisfies specs.

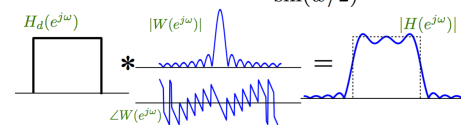
FIR Design by Windowing

- Desired filter,

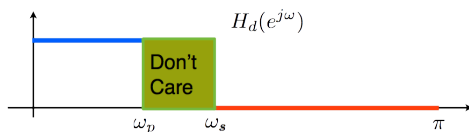
$$H(e^{j\omega}) = H_d(e^{j\omega}) * W(e^{j\omega})$$

- For Boxcar (rectangular) window

$$W(e^{j\omega}) = e^{-j\omega \frac{M}{2}} \frac{\sin(\omega(M+1)/2)}{\sin(\omega/2)}$$



FIR Design by Optimality



- Least Squares:

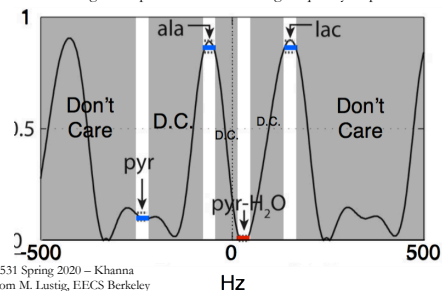
$$\text{minimize} \int_{\omega \in \text{care}} |H(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

- Variation: Weighted Least Squares:

$$\text{minimize} \int_{-\pi}^{\pi} W(\omega) |H(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

Example of Complex Filter

- Larson et. al, "Multiband Excitation Pulses for Hyperpolarized ^{13}C Dynamic Chemical Shift Imaging" JMR 2008;194(1):121-127
- Need to design 11 taps filter with following frequency response:



Admin

- ❑ Find web, get text, start HW 0 and assigned reading...
 - <http://www.seas.upenn.edu/~ese531>
 - <https://piazza.com/upenn/spring2020/ese531/>
 - <https://canvas.upenn.edu/>
- ❑ Diagnostic quiz due 1/23
- ❑ Remaining Questions?