

ESE 531: Digital Signal Processing

Lec 1: January 17, 2019
Introduction and Overview



Lecture Outline

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

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

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

Learning Objectives

- In other words...

□ Math, Math, Math*

*With MATLAB application for intuition

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

Course Structure

- Course Staff (complete info on course website)
- Instructor: Tania Khanna
 - Office hours – Wednesday 2-4 pm or by appointment
 - Email: taniak@seas.upenn.edu
 - Best way to reach me
- TAs:
 - Taishan Li
 - Office hours – WF 10am-11:30am in TBD
 - Mingxuan Sun
 - Office hours – TTh 3-4:30pm in TBD
- Grader: Yulai Weng

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

Course Structure - Assignments/Exams

- Homework – one week long (9 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%]

Course Policies

See web page for full details

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

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

Course Content

Wk	Topic	Lecture	Slides	Due	Reading
1	1/17 Th	Intro/Overview			review course syllabus, complexity
2	1/22 F	Discrete Time Signals & Systems: Part 1			2.1-2.2
3	1/29 Sa	Discrete Time Signals & Systems: Part 2		2/6-6	2.3-2.4
4	1/29 Th	Discrete Time Fourier Transforms			2.5-2.7
5	1/31 Sa	2D Transforms		2/6-6	3.0-3.1
6	2/5 Th	2D Transforms		2/6-6	3.2
7	2/12 Th	Sampling and Reconstruction		2/6-6	4.0-4.3
8	2/19 Th	Discrete Processing of CTDT Signals		2/6-6	4.4-4.5
9	2/26 Th	Discrete Processing of CTDT Signals (cont.), Impulse Invariance		2/6-6	4.6-4.7
10	3/5 Th	Downsampling/Interpolation and Practical Interpolation		2/6-6	4.8-4.9
11	3/12 Th	Non-Integer and Multi-rate Sampling		2/6-6	5.0-5.3
12	3/19 Th	Data Converters and Noise Shaping		2/6-6	5.4-5.5
13	3/26 Th	Frequency Response of LTI Systems		2/6-6	5.6-5.7
14	4/2 Th	Fast Fourier Transform		2/6-6	5.8-5.9
15	4/9 Th	Fast Fourier Transform		2/6-6	6.0-6.1
16	4/16 Th	Fast Fourier Transform		2/6-6	6.2-6.3
17	4/23 Th	Fast Fourier Transform		2/6-6	6.4-6.5
18	4/30 Th	Fast Fourier Transform		2/6-6	6.6-6.7
19	5/7 Th	Fast Fourier Transform		2/6-6	6.8-6.9
20	5/14 Th	Fast Fourier Transform		2/6-6	6.10-6.11
21	5/21 Th	Fast Fourier Transform		2/6-6	6.12-6.13
22	5/28 Th	Fast Fourier Transform		2/6-6	6.14-6.15
23	6/4 Th	Fast Fourier Transform		2/6-6	6.16-6.17
24	6/11 Th	Fast Fourier Transform		2/6-6	6.18-6.19
25	6/18 Th	Fast Fourier Transform		2/6-6	6.20-6.21
26	6/25 Th	Fast Fourier Transform		2/6-6	6.22-6.23
27	7/2 Th	Fast Fourier Transform		2/6-6	6.24-6.25
28	7/9 Th	Fast Fourier Transform		2/6-6	6.26-6.27
29	7/16 Th	Fast Fourier Transform		2/6-6	6.28-6.29
30	7/23 Th	Fast Fourier Transform		2/6-6	6.30-6.31

What is DSP



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Bohemian Rhapsody



□ <https://www.youtube.com/watch?v=I3iIJu08t-Y>

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Bohemian Rhapsody



□ <https://www.youtube.com/watch?v=yE8SYzZ6Eo>

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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,...

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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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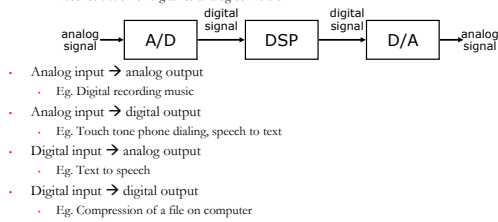
Limitations of Analog Signal Processing

- ❑ 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
- ❑ 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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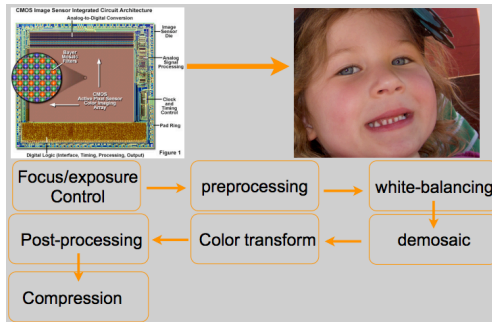
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Historical Forms of Compression

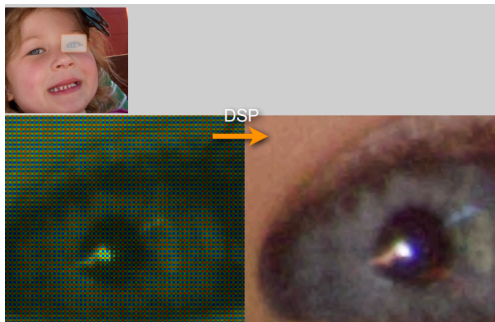
- Morse code: dots (1 unit) dashes (3 units)
 - Code Length inversely proportional to frequency of character
 - E (12.7%) = . (1 unit) Q (0.1%) = --.- (10 units)
- “92 Code”
 - Used by Western-Union in 1859 to reduce BW on telegraph lines by numerical codes for frequently used phrases
 - 1 = wait a minute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

73 Best Regards 19units
 73 Best Regards 59units

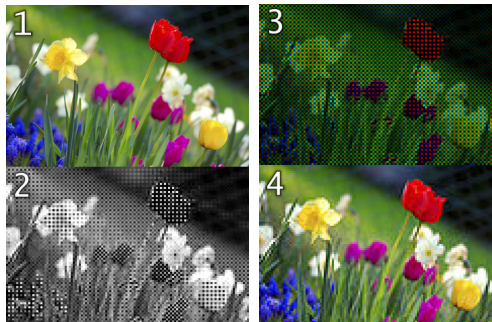
Example II: Digital Imaging Camera



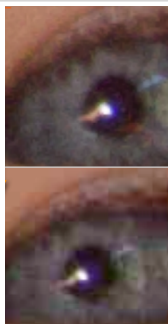
Example II: Digital Imaging Camera



Example II: Digital Imaging Camera

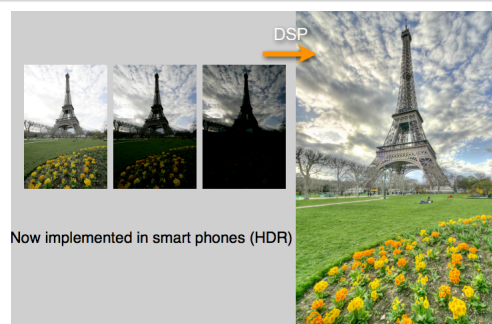


Example II: Digital Imaging Camera



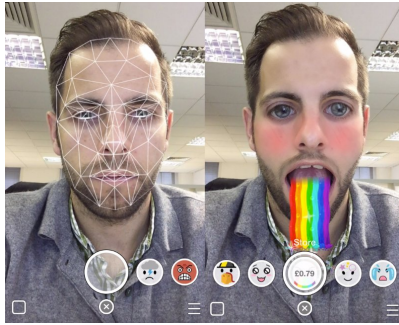
- Compression of 40x without perceptual loss of quality.
- Example of slight overcompression: difference enables 60x compression!

Computational Photography



Now implemented in smart phones (HDR)

Image Processing



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Image Processing - Saves Lives

Canadian 'swirl face'
Thailand

jailed in



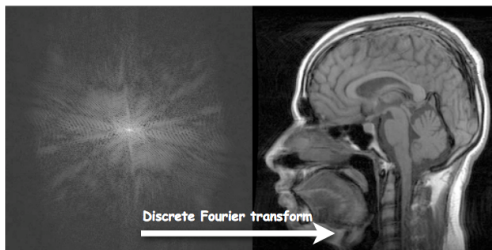
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Example III: MRI

k-space (raw data)

Image

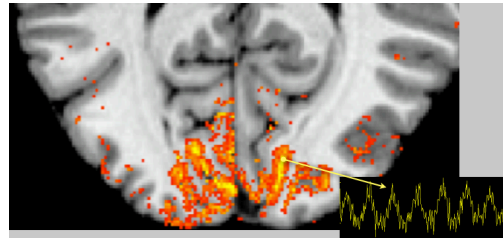


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fMRI example

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

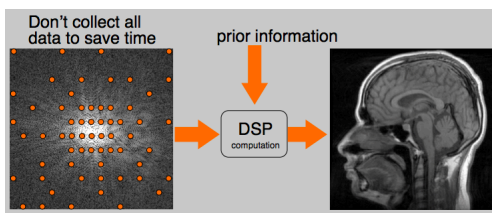


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Compressive Sampling

- Compression meets sampling

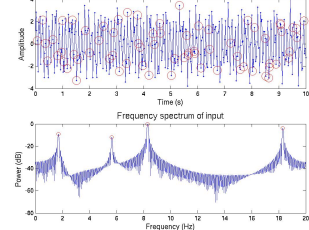


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Example: Sum of Sinusoids

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



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

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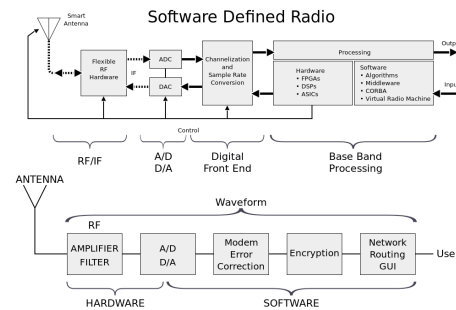
Example IV: Software Defined Radio

- ❑ 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

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Software Defined Radio



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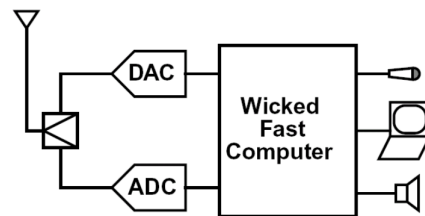
Software Defined Radio

- ❑ 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....

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Software Radio Vision

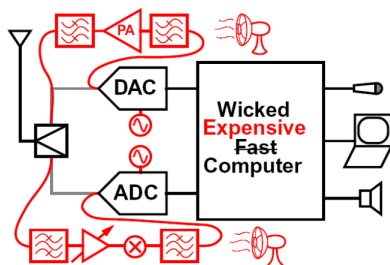


[Schreier, "ADCs and DACs: Marching Towards the Antenna," GIRAFE workshop, ISSCC 2003]

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Software Radio Reality



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

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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

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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.

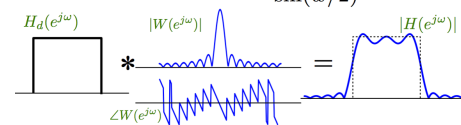
FIR Design by Windowing

- Desired filter,

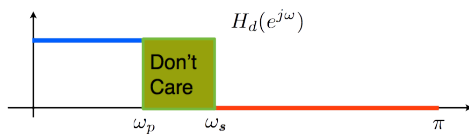
$$H(e^{j\omega}) = H_d(e^{j\omega}) * W(e^{j\omega})$$

- 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:

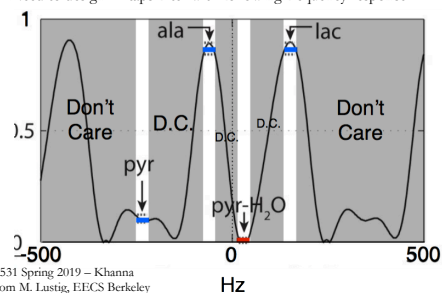
$$\text{minimize} \int_{\omega \in \text{care}} |H(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

- 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/spring2019/ese531/>
 - <https://canvas.upenn.edu/>
- ❑ Remaining Questions?