

ECE 420
Lecture 2
Jan 28 2019

The Compromise of Embedded DSP

- Embedded DSP is an inherent compromise of many, many constraints
 - Beyond just developing an algorithm so that it performs the desired action / produces the desired result
- Platform itself may be fixed a priori, or if a part of the design then cost considerations will limit what can be included
- Limits on
 - Computational capability
 - Memory speed and capacity
 - Word size (floating vs. fixed point)
 - Battery/power

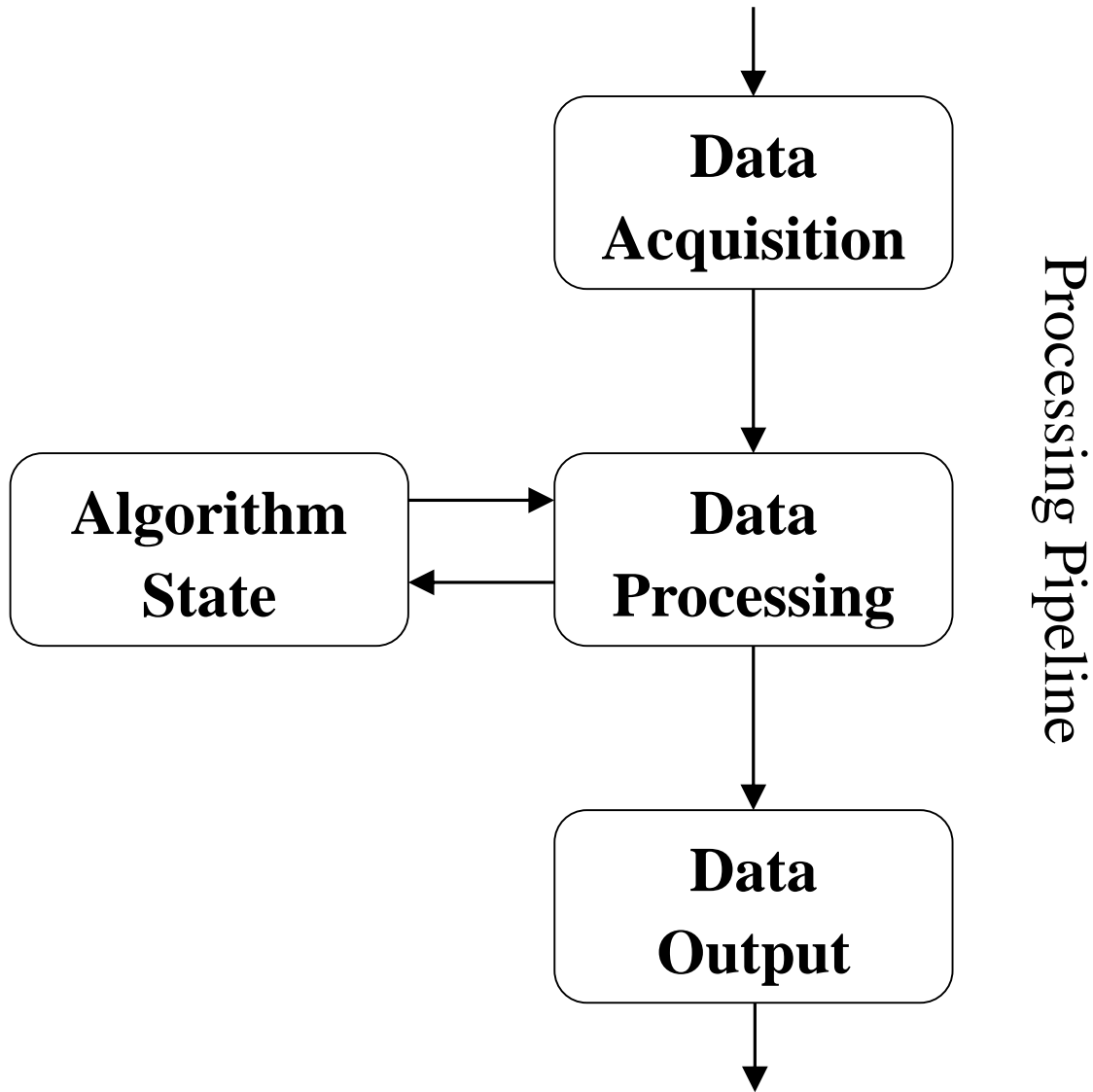
The Compromise of Embedded DSP

- Real-time constraints can pose the biggest design challenges
 - Even “off-line” processing algorithms still need to complete in moderate short order for user satisfaction
- Real-time processing of data requires a certain level of throughput or perceived (and actual) performance will suffer!
 - Once the firehose of data is turned on, the algorithm must be able to process the data quickly enough or bad things will happen
 - Bad things = choppy/missing data in output streams, or malfunctioning algorithms = Unhappy users
- Data rates allow to perform envelope calculations about the target performance of your algorithm

Block Data Processing

- Blocking up the data into batches helps address some of these constraints
- Both practically and theoretically motivated practice
 - We can't hold all the data in memory at once
 - We don't want to wait for all the data to produce output
 - Relative to the processing task at hand, the data tends to become uncorrelated given enough distance in time
- The information we want is available locally in the data
 - How do we get it out? What algorithm to use?

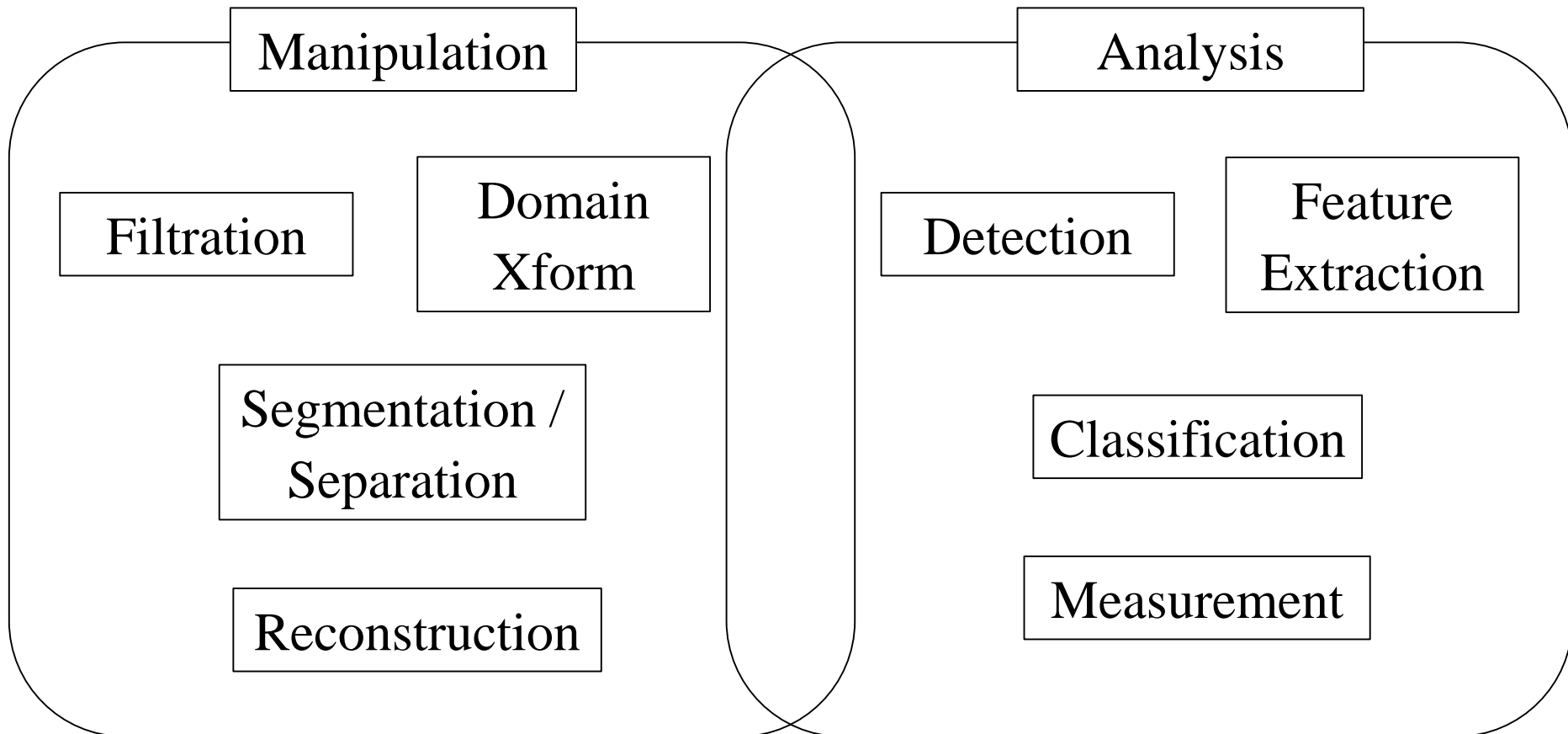
Block Processing



Block Data Processing

- Calculate properties of your algorithm
 - What is a practical/required block size of data?
 - How much computation is required?
 - How much memory is required?
 - Don't forget working buffers and internal state of the algorithm!
 - How much time is allotted?
- Doing these high level calculations will help determine how well an algorithm will map to a particular embedded system
 - An extension of the 'measure twice, cut once' mentality

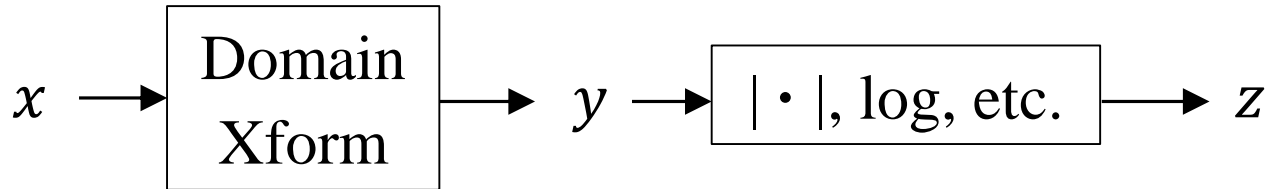
Signal Processing Atlas



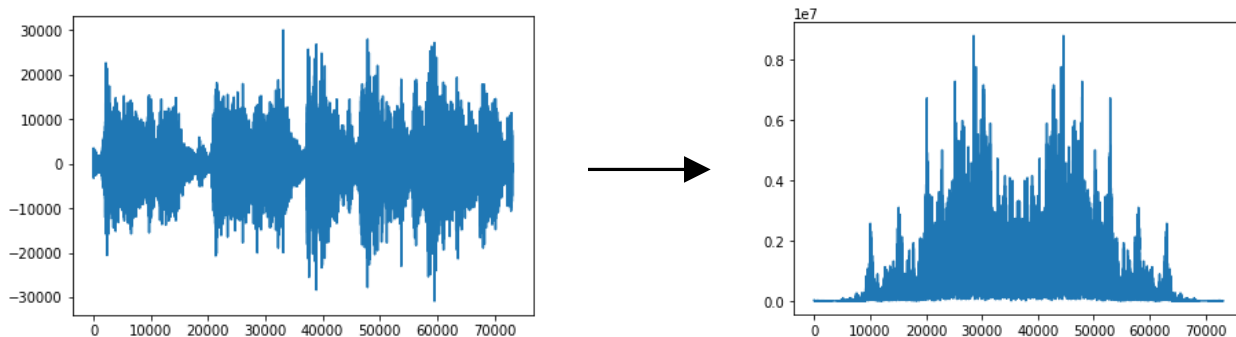
Lab Summary Thus Far

- Lab 1 - Pedometer
 - Detection Algorithm
- Lab 2 - Digital Filtering
 - Filtration
- Lab 3 - Spectrogram [This week]
 - Domain transformation (with a little analysis)

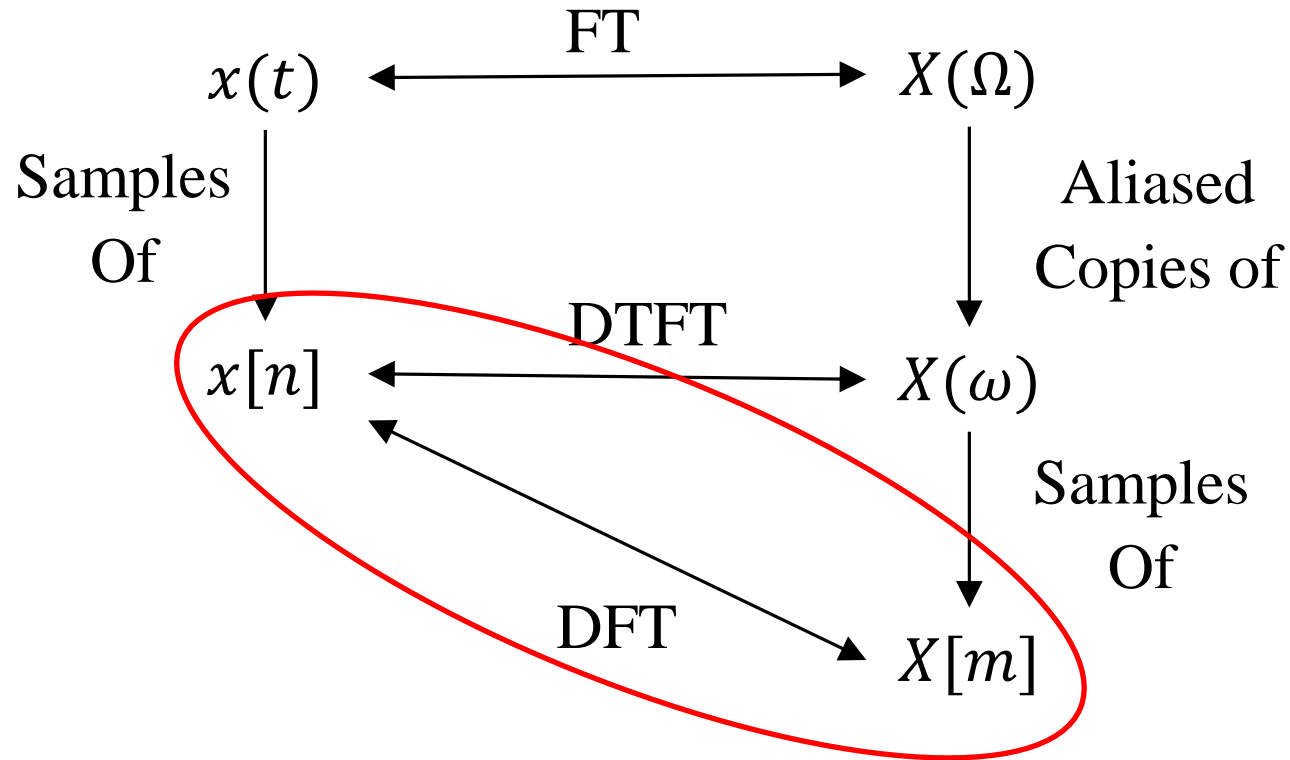
Spectral Analysis



- Relative distribution of signal 'energy' in a different basis
- Common choice is Fourier basis
- Magnitude/log or other post-processing if necessary or for human consumption



Fourier Transforms (An aside)

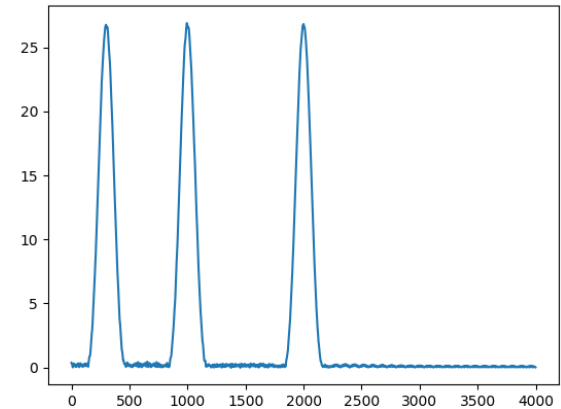
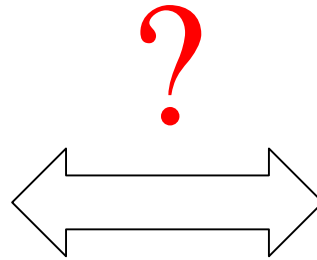
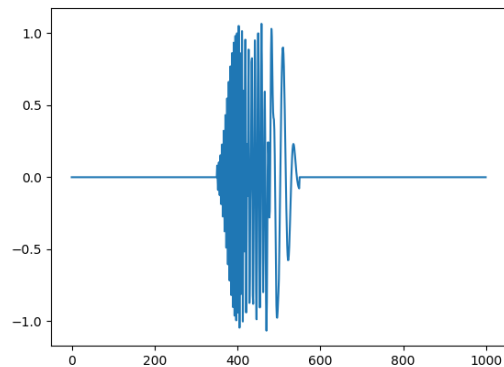
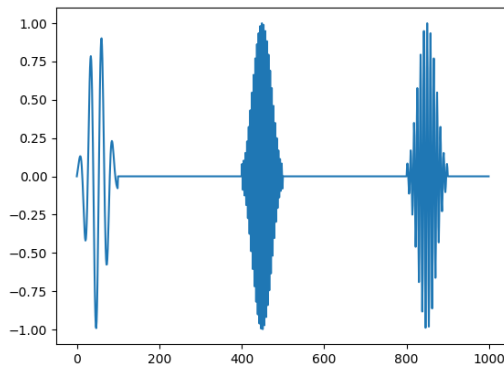
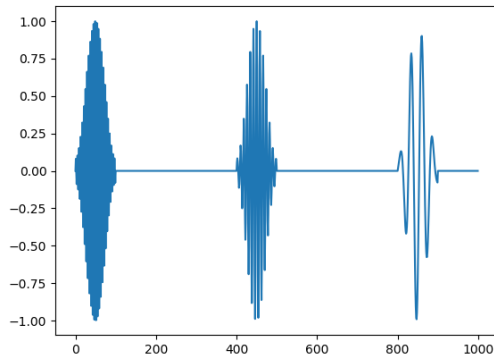


Our good friend the Fast Fourier Transform is which of these?

Digital Fourier Transforms (An aside)

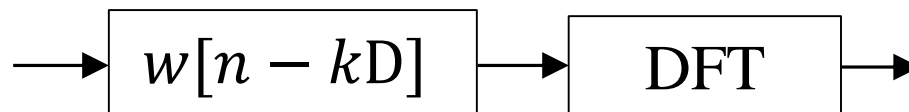
- Two useful properties of DFTs
- Element-wise multiplication in one domain equivalent to circular convolution in the other domain
- Zero padding in one domain equivalent to increased sampling density in the other domain

Fourier Transform Localization



Short Time Fourier Transform

- Divide the signal into (windowed) blocks of data
- Apply DFT to each block
 - Take magnitude if desired
- Solves* our local spectral analysis problem



A Brief Word from Allen77

- Continuous STFT

- $X(f, t) = \int_{-\infty}^{\infty} w(t - \tau)x(\tau)e^{j2\pi f\tau} d\tau$

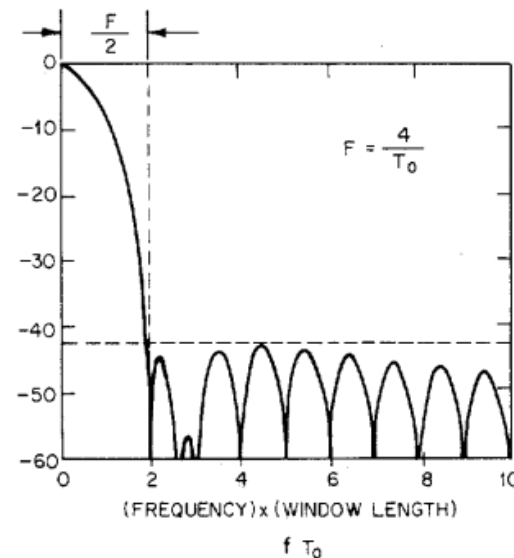
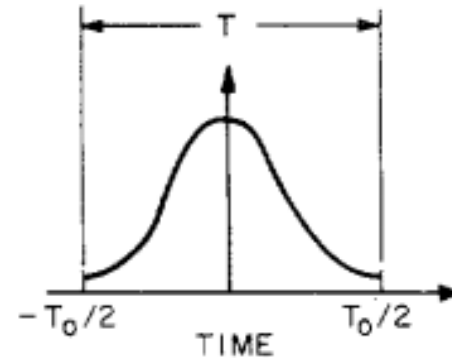
- Assume window $w(t)$ is time limited and essentially bandlimited function

- Questions to be answered

- What sampling is required in f, t ?
 - How does the choice of $w(t)$ influence this?
 - How can this STFT be used as an analysis/synthesis filterbank?

Discretization of STFT

- Window function has effective time/frequency limits of T, F respectively
 - For Hamming window, $F = 4/T$
- Allen argues sampling of
 - $t_n = nD = n/F$
 - $f_m = m/T$
- Discrete STFT
 - $X_{nm} = DFT\{w[nD - k]x[k]\}$



Synthesis Properties of STFT

- Up to this point, discussed the *Analysis* properties of STFT
- What about *Synthesis* properties?
- Exploiting the property that $w[k]$ forms a partition of unity

$$\sum_{n=-\infty}^{\infty} w[nD - k] = 1$$

- The signal is directly recoverable

$$x[k] = \sum_n DFT^{-1}\{X_{nm}\}$$

Using Synthesis for Signal Manipulation

- Spectral-based filtering operations can be carried out in STFT domain
- Leverage Fourier multiplication == spatial convolution
 - Requires zero padding to prevent circular aliasing!
- Can implement Fourier based filtering routines using STFT when manipulation is a shift invariant filter
- Can also implement spatially varying filtration at your own risk

Limitations of STFT

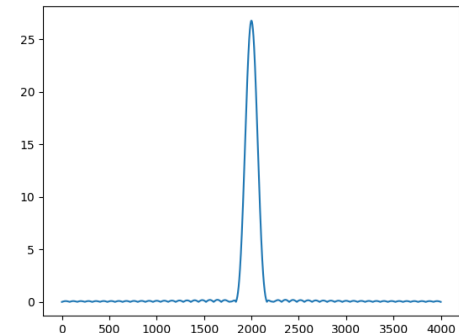
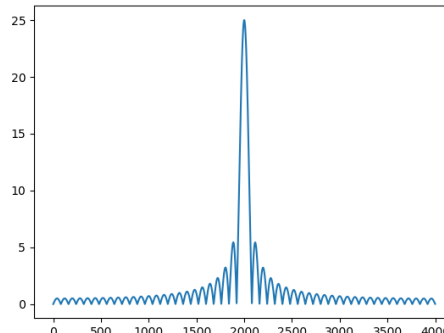
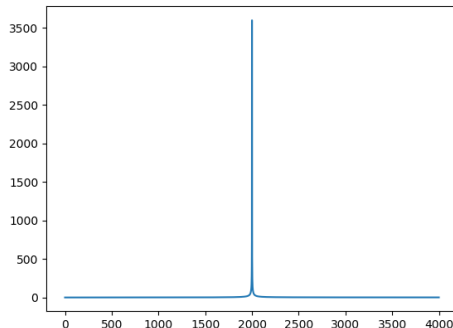
- When using overlapping windows, straightforward synthesis is possible
- This comes at the expense of extra data
 - Expansion of data = $D/T = TF$
 - Allen claims this allows for more stable synthesis when spatially varying filters are considered
- Time/frequency localization tradeoff not eliminated
 - Short $T \rightarrow$ Wide F and vice-versa

Signal Detection / Separation

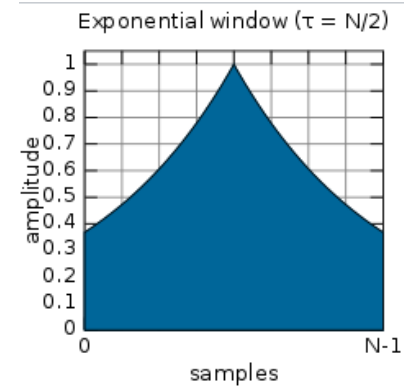
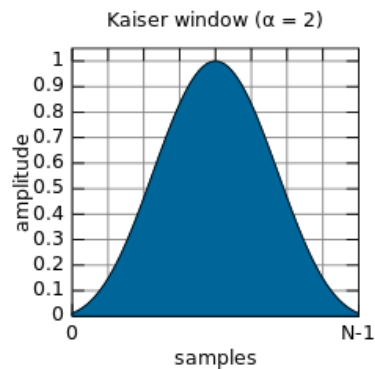
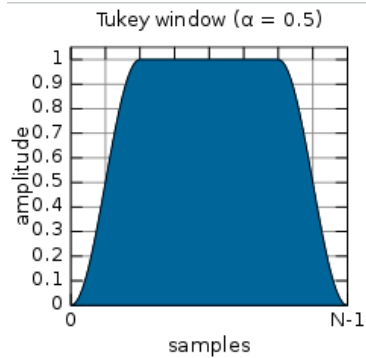
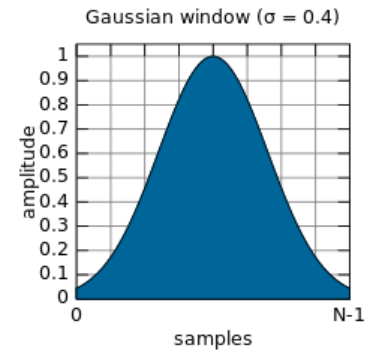
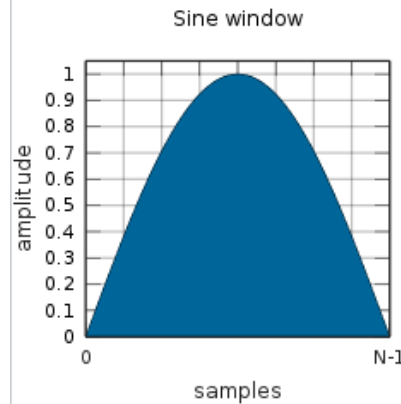
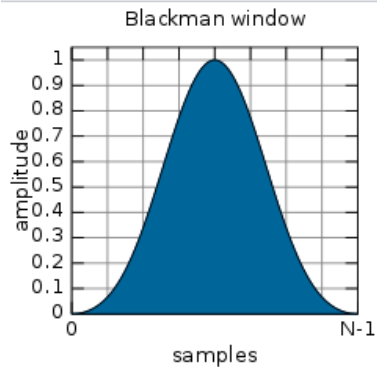
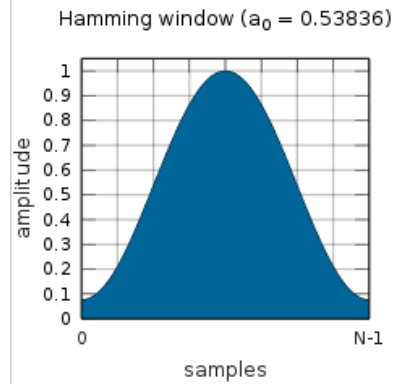
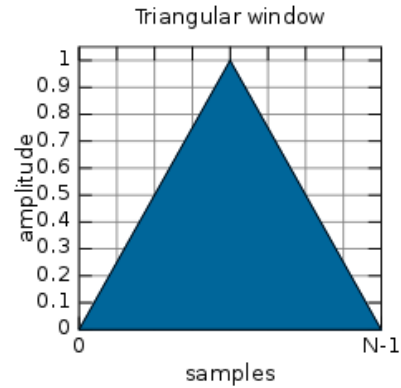
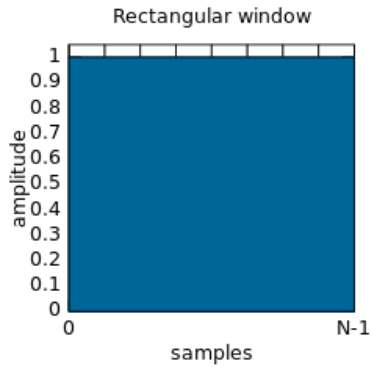
- Problem statement: we want to identify key components of a signal
- Example: Finding peaks in our spectrogram
- Question: Where are the peak locations and intensities?
- Challenging Scenarios:
 - Peaks closely spaced to one another
 - Peaks with relatively small amplitude
 - Peaks with relatively small amplitude spaced closely to one another

Effect of Windowing on Spectral Analysis

- Just by blocking the data we are applying a window
- DFT tells us windowing results in convolution in frequency domain
 - Signals that might once had good frequency localization become muddled
 - This convolution can be viewed as the response of our analysis system and characterizes the performance of the ultimate algorithm
- Why even care about windowing?



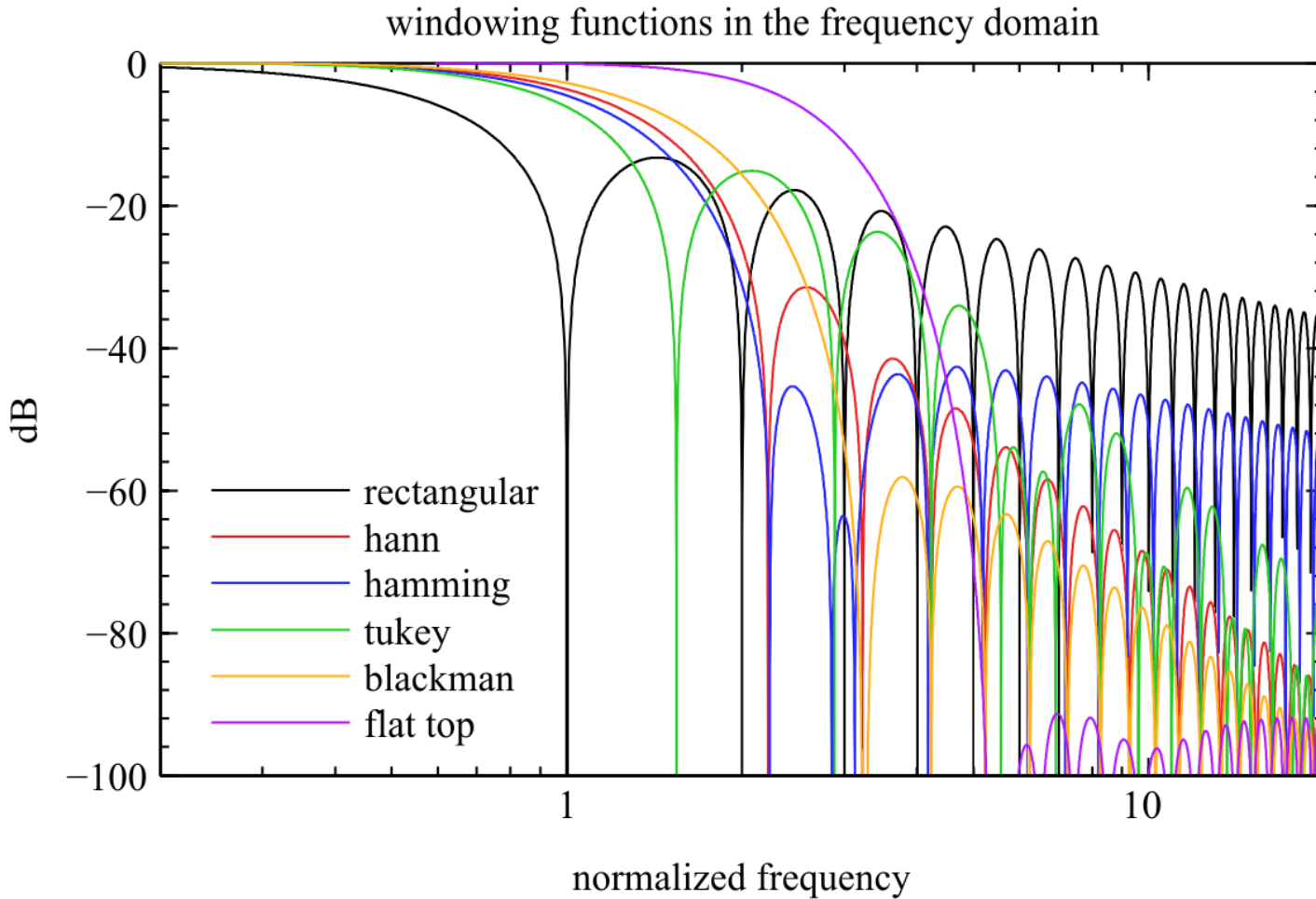
Window Selection



Window Selection

- Windows generally provide a tradeoff between mainlobe width and sidelobe height (or sidelobe decay)
 - Determines how well we can resolve neighboring signals
- Generally a good idea to select a window with an effective bandlimit
 - This is why not explicitly applying a window is not a good strategy
- From a detection performance perspective, we want the transform to be as close to a delta as possible
 - In practice we have to decide on the tradeoff

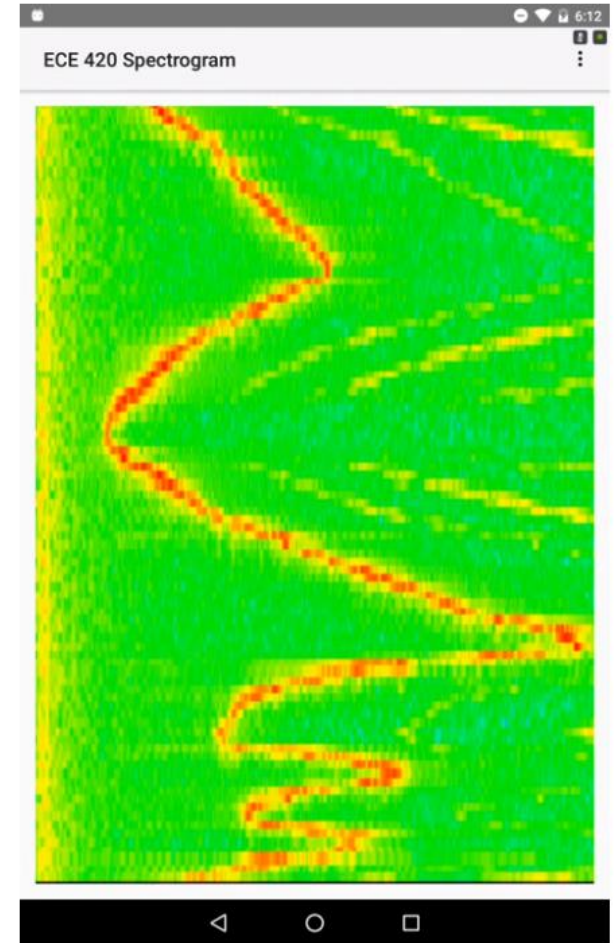
Window Comparison



By Aleks.labuda - This diagram was created with MATLAB., CC BY-SA 3.0
<https://commons.wikimedia.org/w/index.php?curid=18933145>

Real Time Spectral Analyzer

- Block processing and analysis on the tablet using STFT
- Move to floating point data processing
- Prize for fastest implementation of the FFT algorithm
- (KIDDING)
- ALWAYS leverage existing libraries unless explicitly told not to
 - Correct
 - Well tuned/optimized



This week

- Lab 2: Digital Filtering Quiz/Demo
- Lab 3: Real time spectral analyzer
- Be thinking about Assigned Project Labs
 - More info at lecture next week