



Signal & Systems

Lecture 9: Windowing

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Recap: Spectral Leakage

Key idea

- In practice, we observe only a **finite-length** segment of a signal.
- This is equivalent to multiplying the signal by a **window**:

$$x_w[n] = x[n] w[n]$$

- With a rectangular observation window,

$$w[n] = \begin{cases} 1, & 0 \leq n \leq N - 1 \\ 0, & \text{otherwise} \end{cases}$$

Frequency-domain consequence

- Multiplication in time \iff convolution in frequency
- A pure sinusoid no longer appears as a single spectral line
- Its energy **spreads into nearby frequencies**:

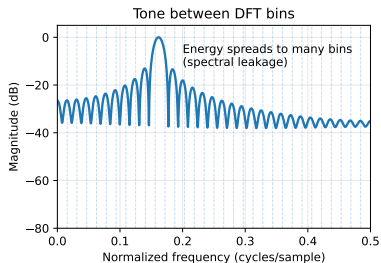
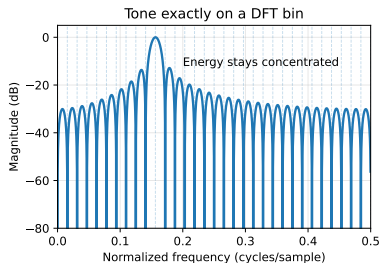
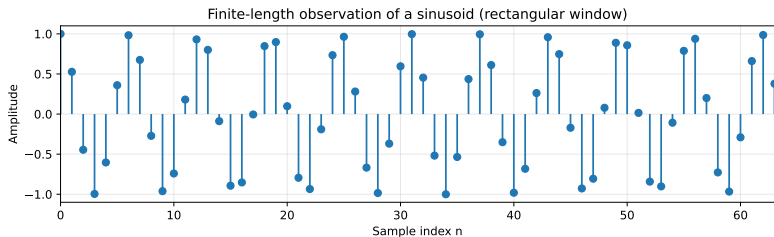
$$X_w(e^{j\omega}) = X(e^{j\omega}) * W(e^{j\omega})$$

Interpretation

- Sharp truncation at the edges creates spectral spreading
- This spreading is called **spectral leakage**

Recap: Spectral Leakage (2)

Spectral Leakage from Finite Observation



Definition

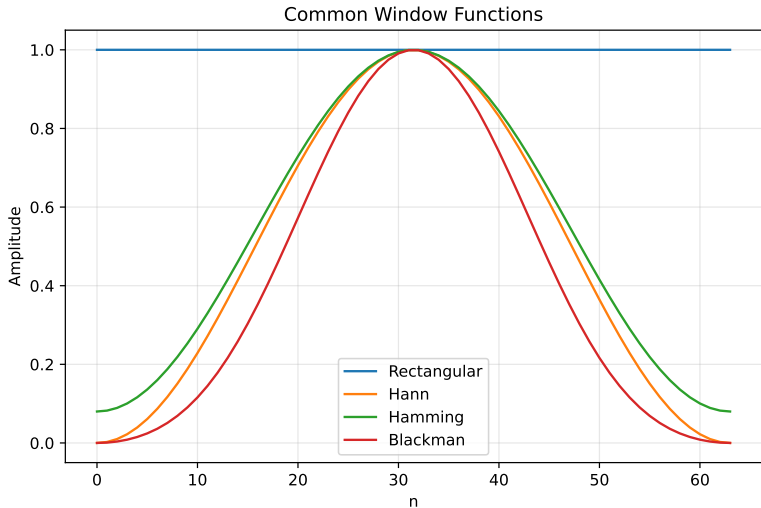
- Windowing means multiplying a signal by a function that selects and shapes a segment:

$$x_w[n] = x[n] w[n]$$

Interpretation

- We never observe infinite signals in practice
- We **choose how to observe** the signal
- The window $w[n]$ defines:
 - what part we keep
 - how smoothly we taper the edges

Window Shapes Example



Original signal

- Potentially long or infinite
- May not align with observation boundaries

After windowing

- Signal is **truncated and shaped**
- Edges become important

Key intuition

- Rectangular window:
 - sharp discontinuities
- Smooth windows:
 - gradual transitions
 - fewer abrupt jumps

Original signal

- Potentially long or infinite
- May not align with observation boundaries

After windowing

- Signal is **truncated and shaped**
- Edges become important

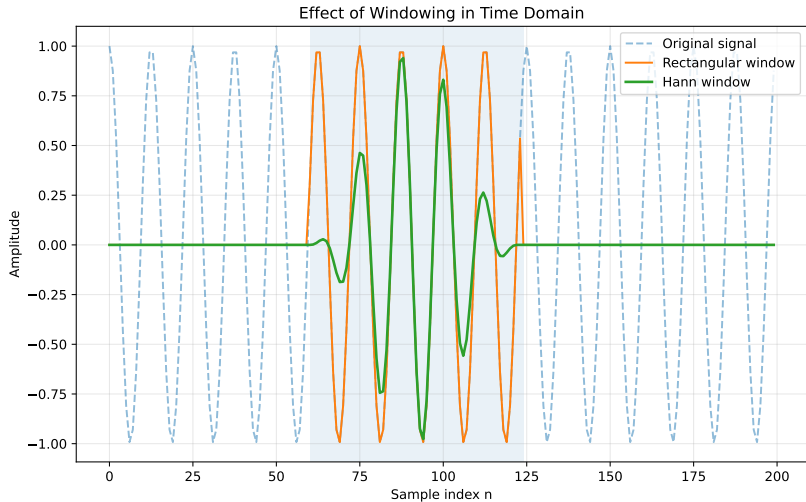
Key intuition

- Rectangular window:
 - sharp discontinuities
- Smooth windows:
 - gradual transitions
 - fewer abrupt jumps

Why this matters

- Discontinuities → high-frequency components

Effect of Windowing in Time (2)



Fundamental relation

- Multiplication in time \iff convolution in frequency

$$X_w(e^{j\omega}) = X(e^{j\omega}) * W(e^{j\omega})$$

Interpretation

- The spectrum of the signal is **smear**ed by the window
- The window acts as a **spectral filter**

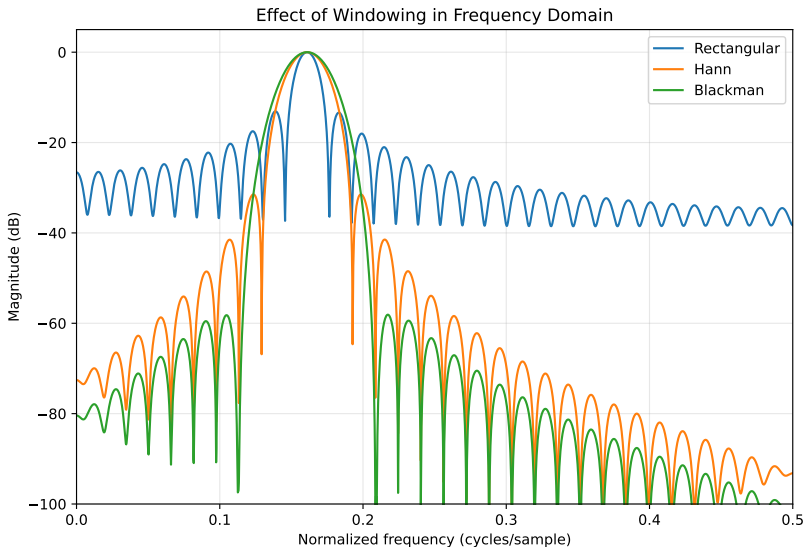
Key consequence

- Rectangular window \rightarrow strong side-lobes (more leakage)
- Smooth window \rightarrow lower side-lobes (less leakage)

Big idea

- You do **not** just observe the signal
- You observe the signal *through the window*

Effect of Windowing in Frequency (2)



By choosing a window, you control:

- Spectral leakage (side-lobes)
- Frequency resolution (main-lobe width)
- Dynamic range (ability to see weak signals)

Key intuition

- Sharp window → narrow main lobe, high side-lobes
- Smooth window → wide main lobe, low side-lobes

Takeaway

Windowing is not just preprocessing — it is a **design choice** that determines what you can see in the spectrum.

The Fundamental Tradeoff

Choosing a window means balancing two competing goals:

- 1 **Narrow main lobe** → good **frequency resolution**
- 2 **Low side lobes** → low **spectral leakage**

Visual intuition

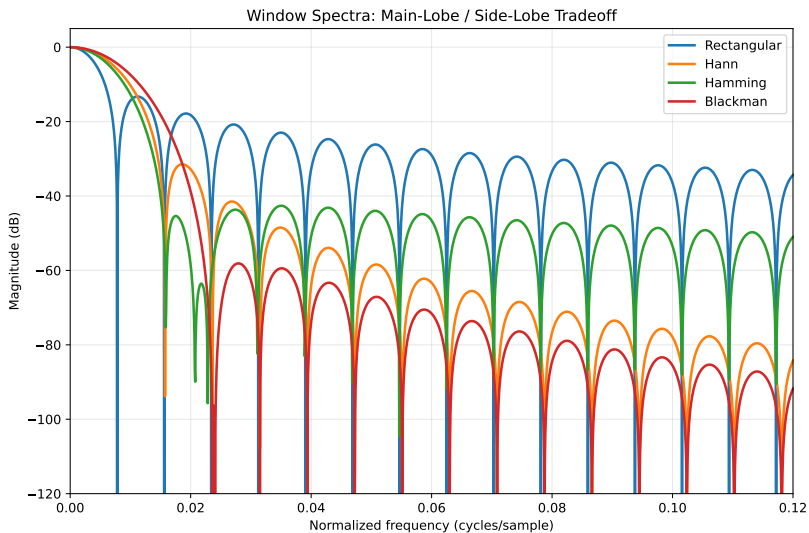
- **Main lobe** = how precisely we can locate a frequency
- **Side lobes** = how much energy spills into nearby frequencies

Key point

- We cannot optimize both simultaneously
- A sharper window in time gives:
 - narrower main lobe
 - higher side lobes
- A smoother window in time gives:
 - lower side lobes
 - wider main lobe

Windowing is a controlled tradeoff.

The Fundamental Tradeoff (2)



Why Different Windows Behave Differently

Fundamental relation

$$x_w[n] = x[n] w[n]$$
$$X_w(e^{j\omega}) = X(e^{j\omega}) * W(e^{j\omega})$$

Interpretation

- Windowing in time means **convolution in frequency**
- Each window has its own **spectral signature**
- The observed spectrum is the original spectrum blurred by the window spectrum

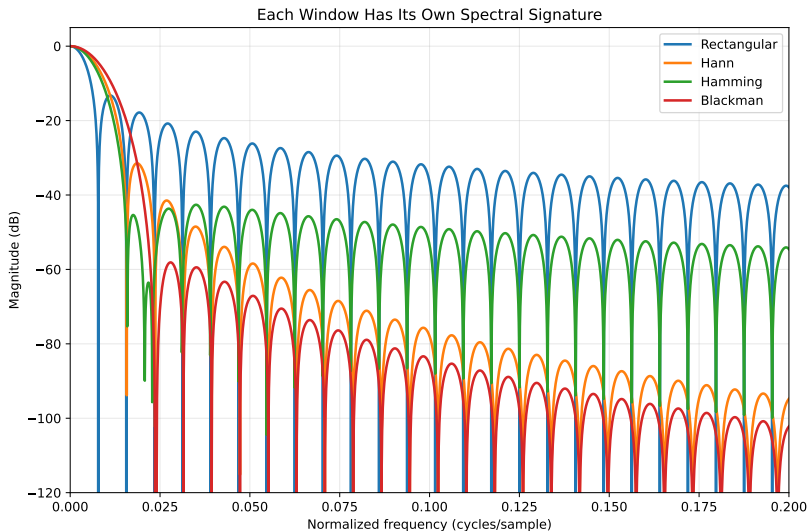
Consequences

- Narrow spectral kernel \rightarrow better resolution
- Low side-lobe kernel \rightarrow less leakage
- Different windows behave differently because $W(e^{j\omega})$ is different

Takeaway

You do not observe the signal alone; you observe the signal *through the spectrum of the window*.

Why Different Windows Behave Differently (2)



Common Window Definitions I

For $0 \leq n \leq N - 1$, the windowed signal is

$$x_w[n] = x[n] w[n].$$

Rectangular window

$$w_{\text{rect}}[n] = 1$$

Hann window

$$w_{\text{Hann}}[n] = \frac{1}{2} \left(1 - \cos \left(\frac{2\pi n}{N-1} \right) \right)$$

Hamming window

$$w_{\text{Hamming}}[n] = 0.54 - 0.46 \cos \left(\frac{2\pi n}{N-1} \right)$$

Observation

Rectangular keeps all samples equally. Hann and Hamming taper the edges smoothly toward zero or near-zero.

Common Window Definitions II

For $0 \leq n \leq N - 1$:

Blackman window

$$w_{\text{Blackman}}[n] = 0.42 - 0.5 \cos\left(\frac{2\pi n}{N-1}\right) + 0.08 \cos\left(\frac{4\pi n}{N-1}\right)$$

Kaiser window

$$w_{\text{Kaiser}}[n] = \frac{I_0\left(\beta \sqrt{1 - \left(\frac{2n}{N-1} - 1\right)^2}\right)}{I_0(\beta)}$$

where $I_0(\cdot)$ is the modified Bessel function of the first kind.

- $\beta = 0$ gives a rectangular-like shape
- Larger $\beta \rightarrow$ smoother taper, lower side-lobes, wider main lobe

Why Kaiser is useful

It gives a tunable tradeoff between resolution and leakage.

Many Windows Are Cosine Sums

Many practical windows can be written as

$$w[n] = \sum_{m=0}^M a_m \cos\left(\frac{2\pi mn}{N-1}\right), \quad 0 \leq n \leq N-1.$$

Examples:

- Rectangular: only a constant term
- Hann: constant + one cosine term
- Hamming: constant + one cosine term with different coefficients
- Blackman: constant + two cosine terms

Interpretation

More shaping freedom in time \rightarrow more control over side-lobes and main-lobe width in frequency.

Comparing Common Windows

Window	Main Lobe Width	Side Lobes	Dynamic Range	Typical Use
Rectangular	Narrowest	Worst leakage	Poor	Frequency estimation for clean, well-aligned tones
Hann	Wider	Much lower	Good	General-purpose spectral analysis
Hamming	Similar to Hann	Slightly lower first side-lobe	Good–better	Audio / practical FFT analysis
Blackman	Wide	Very low	High	Weak tones near strong interferers
Kaiser	Adjustable	Adjustable	Adjustable	Tunable engineering compromise

Key takeaway

Windows move you along the resolution–leakage–dynamic-range tradeoff curve.

Punchline

Your window determines your usable dynamic range.

Why?

- A strong sinusoid produces side-lobes
- These side-lobes act like an **artificial floor**
- A weak sinusoid must rise above that floor to be visible

Interpretation

- High side-lobes → poor dynamic range
- Low side-lobes → high dynamic range

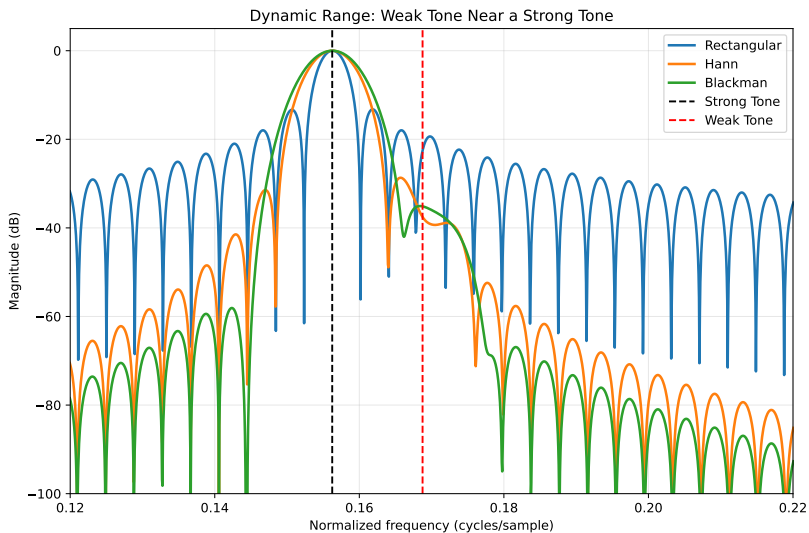
Visual intuition

- **Main lobe** tells us how well we separate nearby tones
- **Side lobes** tell us whether weak tones get buried

Important

- Better dynamic range usually comes with a wider main lobe
- So improving detectability often reduces resolution

Windowing and Usable Dynamic Range (2)



Key idea

- A strong sinusoid does not remain perfectly localized in the spectrum
- Because of the window, it creates a **main lobe** and **side lobes**
- The side lobes form an **artificial floor** around the tone

Interpretation

- If a weak tone lies below that floor, it becomes hard to detect
- Therefore, side-lobe level directly affects **usable dynamic range**

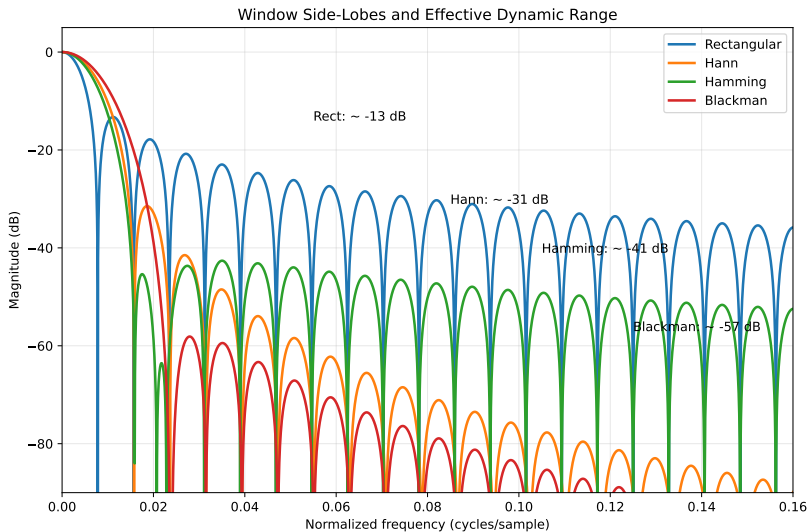
Question to ask

Can I see a signal that is 40 dB below a stronger one?

Rule of thumb

- Higher side lobes → poorer dynamic range
- Lower side lobes → better weak-signal visibility

Side-Lobes as an Artificial Floor (2)



Approximate Side-Lobe Floors of Common Windows

Window	Main Lobe	First Side-Lobe Level	Dynamic Range Intuition
Rectangular	Narrow	≈ -13 dB	Poor
Hann	Medium	≈ -31 dB	Good
Hamming	Medium	≈ -41 dB	Better
Blackman	Wide	≈ -57 dB	Excellent

Interpretation

A window with a side-lobe level near -40 dB is much more suitable for revealing tones that are 30 to 40 dB below a strong one.

Windows with lower side lobes provide higher usable dynamic range.

How Do We Compute the Side-Lobe Level?

Let $w[n]$ be a window of length N , and let

$$W[k] = \sum_{n=0}^{N-1} w[n] e^{-j2\pi kn/N_{\text{FFT}}}$$

be a densely sampled DFT of the window.

Step 1: Normalize by the peak

$$\tilde{W}[k] = \frac{|W[k]|}{\max_m |W[m]|}$$

Step 2: Convert to dB

$$W_{\text{dB}}[k] = 20 \log_{10}(\tilde{W}[k])$$

Step 3: Find the largest side-lobe peak

- Exclude the main-lobe region around $k = 0$
- Search for the highest local maximum outside the main lobe

Definition

$$\text{Side-lobe level} = \max_{k \in \text{side-lobes}} W_{\text{dB}}[k]$$

This value is usually reported relative to the main-lobe peak, in dB.

Worked Example: Hann Window

For the Hann window,

$$w[n] = \frac{1}{2} \left(1 - \cos \left(\frac{2\pi n}{N-1} \right) \right), \quad 0 \leq n \leq N-1$$

Procedure

- 1 Compute a high-resolution FFT of $w[n]$
- 2 Normalize so the peak is at 0 dB
- 3 Ignore the main lobe near DC
- 4 Measure the highest remaining peak

Result

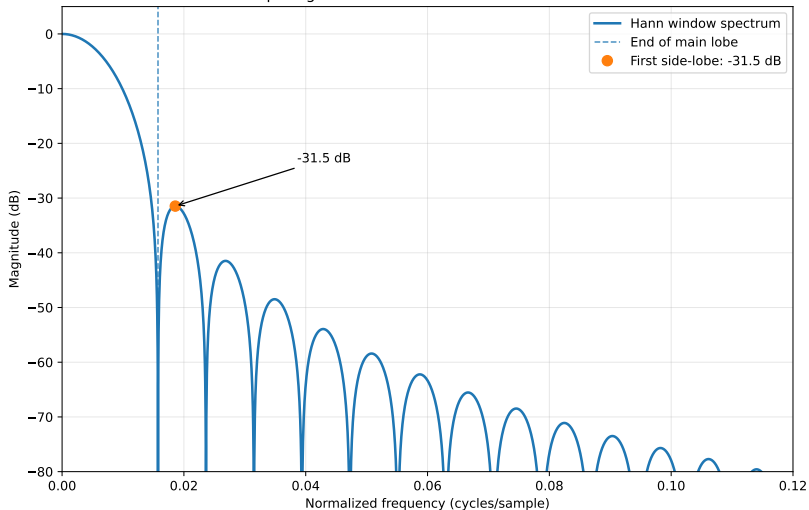
First side-lobe level ≈ -31 dB

Meaning

A strong tone observed with a Hann window creates leakage about 31 dB below its peak in the first side-lobe.

Worked Example: Hann Window (2)

Computing the First Side-Lobe Level: Hann Window



Analytical Example: Side-Lobe of the Rectangular Window

For the rectangular window

$$w[n] = 1, \quad 0 \leq n \leq N - 1,$$

its DTFT is

$$W(e^{j\omega}) = \sum_{n=0}^{N-1} e^{-j\omega n} = e^{-j\omega(N-1)/2} \frac{\sin(N\omega/2)}{\sin(\omega/2)}.$$

Ignoring the linear phase term, the normalized magnitude is

$$\frac{|W(e^{j\omega})|}{N} = \left| \frac{\sin(N\omega/2)}{N \sin(\omega/2)} \right|.$$

For large N , define $x = \frac{N\omega}{2}$, so that near the main lobe / first side-lobe region,

$$\frac{|W(e^{j\omega})|}{N} \approx \left| \frac{\sin x}{x} \right|.$$

The side-lobe peaks occur where

$$\frac{d}{dx} \left(\frac{\sin x}{x} \right) = 0 \implies \frac{x \cos x - \sin x}{x^2} = 0 \implies x \cos x - \sin x = 0 \implies \tan x = x.$$

The first solution after $x = \pi$ is

$$x_1 \approx 4.493.$$

Therefore, the first side-lobe amplitude is $\left| \frac{\sin x_1}{x_1} \right| \approx 0.217$, and its level relative to the main-lobe peak is

$$20 \log_{10}(0.217) \approx -13.26 \text{ dB}.$$

Summary: What a Window Really Does

- Windowing changes the **spectrum of the observation**
- Every window imposes a different balance between:
 - **Resolution** (main-lobe width)
 - **Leakage suppression** (side-lobe level)
 - **Dynamic range** (visibility of weak tones)
- No single window is best for all tasks

Rule of thumb

- Need sharp separation of close tones? → narrower main lobe
- Need to reveal weak tones near strong ones? → lower side lobes

Windowing Changes Amplitude

Observation

- When we apply a window, measured amplitudes decrease
- The same sinusoid gives different peak values for different windows

Why does this happen?

- Windowing multiplies the signal:

$$x_w[n] = x[n] w[n]$$

- Most windows have values less than 1
- So the signal energy is **attenuated**

Key idea

Windowing does not just reshape the spectrum — it also scales the amplitude.

Question

- Can we correct for this amplitude loss?

Definition

The **coherent gain (CG)** of a window is

$$\text{CG} = \frac{1}{N} \sum_{n=0}^{N-1} w[n].$$

Interpretation

- It is the **average value** of the window
- It tells us how much a **coherent sinusoid** is scaled

Special case

- Rectangular window:

$$\text{CG} = 1$$

Meaning

A sinusoid aligned with a DFT bin is scaled approximately by the coherent gain.

Measured vs true amplitude

- Let A_{true} be the true amplitude
- Let A_{measured} be what we observe after windowing

Relation

$$A_{\text{measured}} \approx \text{CG} \cdot A_{\text{true}}$$

Correction

$$A_{\text{true}} \approx \frac{A_{\text{measured}}}{\text{CG}}$$

Practical rule

Always divide by the coherent gain when estimating amplitudes from a windowed FFT.

In dB

$$A_{\text{dB, corrected}} = A_{\text{dB}} - 20 \log_{10}(\text{CG})$$

Typical Coherent Gain Values

Window	Coherent Gain
Rectangular	1.00
Hann	0.50
Hamming	0.54
Blackman	0.42

Interpretation

- Hann reduces amplitude by about 50%
- Blackman reduces amplitude even more

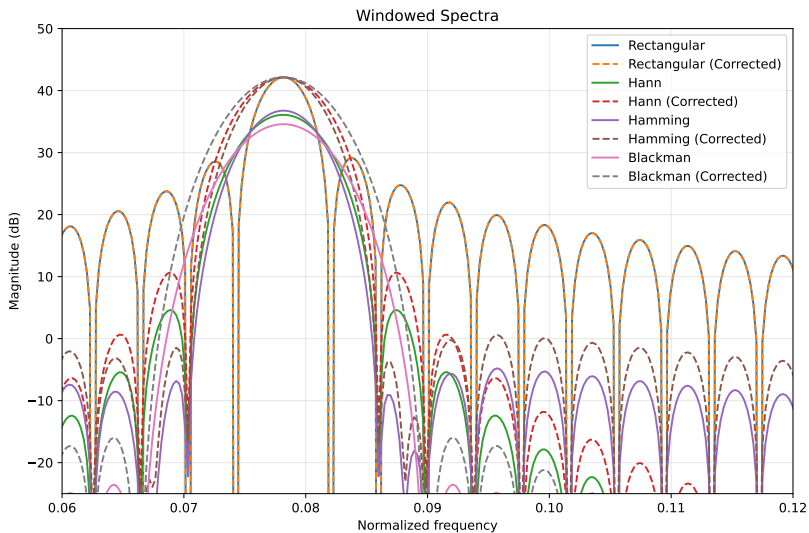
Tradeoff reminder

- Better leakage suppression → lower coherent gain

Takeaway

Windows that improve dynamic range also attenuate amplitude more.

Coherent Gain Example



Experiment: Same Signal, Different Windows

Goal

- Compare how different windows affect what we can observe in the spectrum

Signal

$$x[n] = A_1 \cos(2\pi f_1 n) + A_2 \cos(2\pi f_2 n), \quad A_2 \ll A_1$$

with:

- a **strong** tone at frequency f_1
- a nearby **weak** tone at frequency f_2

Question

- Can we see the weak tone in the presence of the strong one?

What we will compare

Rectangular vs Hann vs Blackman

What Different Windows Reveal

Rectangular window

- Narrow main lobe
- High side-lobes
- Leakage from the strong tone can hide the weak tone

Hann window

- Lower side-lobes
- Weak tone becomes visible
- Main lobe is wider

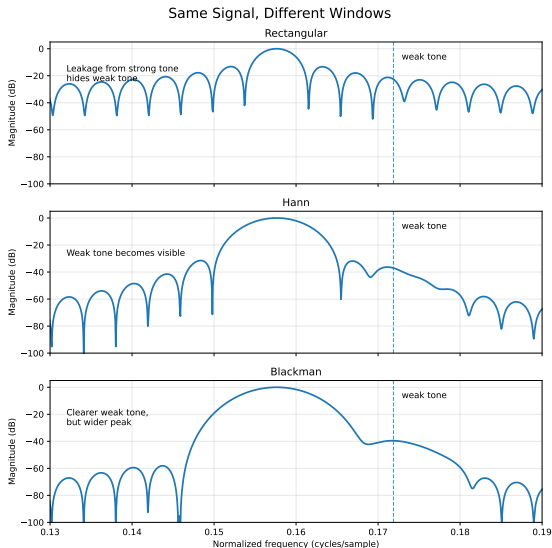
Blackman window

- Very low side-lobes
- Weak tone is very clear
- Peaks are more blurred due to wider main lobe

Key lesson

Window choice determines what you can see.

What Different Windows Reveal (2)



Goal

- Study how window choice affects the ability to separate nearby tones

Signal

$$x[n] = \cos(2\pi f_1 n) + \cos(2\pi f_2 n), \quad f_1 \approx f_2$$

with:

- two **equal-amplitude** sinusoids
- very **small frequency separation**

Question

- Can the spectrum resolve the two tones as distinct peaks?

What we compare

Rectangular window vs Hann window

Rectangular window

- Narrower main lobe
- Better frequency resolution
- Two close tones can still appear as separate peaks
- But the spectrum is more affected by leakage / ripple

Hann window

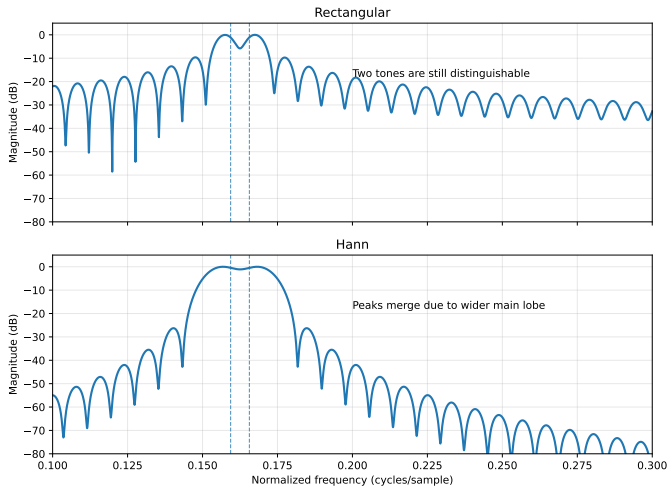
- Lower side-lobes
- Smoother spectrum
- Wider main lobe
- Very close tones may merge into one peak

Key lesson

Reducing leakage usually comes at the cost of frequency resolution.

Experiment: Resolution Tradeoff (2)

Experiment 2: Close Frequencies



Experiment 3: Two-Tone Dynamic Range Test

Goal

- Test how well different windows reveal a weak tone near a strong one

Signal

$$x[n] = A_1 \cos(2\pi f_1 n) + A_2 \cos(2\pi f_2 n)$$

with:

- a strong tone at 0 dB
- a weak nearby tone at:

$$-20 \text{ dB}, \quad -40 \text{ dB}, \quad -60 \text{ dB}$$

Question

- As the weak tone gets smaller, which windows still reveal it?

What we compare

Rectangular vs Hann vs Blackman

Recall

- Rectangular window has narrow main lobe but high side-lobes
- Hann has lower side-lobes
- Blackman has very low side-lobes

Expected behavior

- **Rectangular:** weak tone disappears quickly
- **Hann:** visible up to roughly 30–40 dB difference
- **Blackman:** visible even near 60 dB difference

What determines visibility?

The weak tone must rise above the side-lobe floor created by the strong tone.

What happens as the weak tone decreases?

Rectangular

- High side-lobes from the strong tone
- Weak tone is quickly buried

Hann

- Lower leakage floor
- Weak tone remains visible longer

Blackman

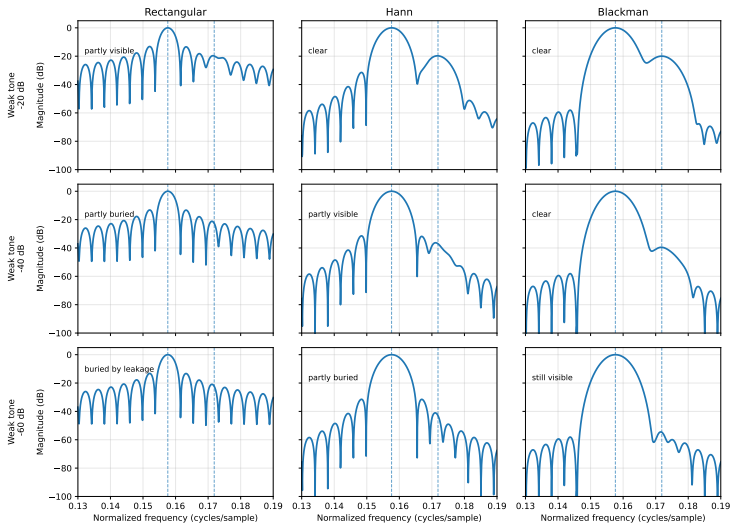
- Very low side-lobes
- Weak tone can remain visible even at very low levels

Key lesson

Dynamic range is not only a property of the signal – it is also a property of the window.

Visualization

Experiment 3: Weak-Tone Visibility vs Window Choice



Did the signal disappear
– or did our tool fail to see it?

Did the signal disappear – or did our tool fail to see it?

- The signal is still there in all cases
- What changes is the **analysis window**
- Different windows create different side-lobe floors
- That floor determines whether weak components remain visible

Takeaway

A spectrum analyzer does not simply reveal truth; it reveals truth through a chosen window.

- **Any Questions?**
- **Office Hours:**
 - **Mon & Tue** (09:00-11:00)
 - 24/7 by email (costashatz@upatras.gr, subject: *ECE_SS_AM*)
- **Material and Announcements**



Laboratory of Automation & Robotics