

Ten Standard Tasks for Evaluating Signal Analysis Code Generation

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Abstract

This document outlines ten standard tasks commonly used in digital signal processing (DSP) and signal analysis. They serve as benchmarks for evaluating whether a large language model can reliably generate code suitable for practical scenarios.

1 Introduction

Signal analysis often involves transforming or filtering data in order to glean insight about underlying structures or behaviors. When assessing a code-generating model for signal analysis, it is valuable to test a broad range of tasks. Below are ten tasks that collectively cover frequency-domain analysis, time-domain filtering, multi-rate operations, time-frequency representations, and more.

2 Tasks Overview

1. Fast Fourier Transform (FFT) of a Given Signal

- *Rationale:* Frequency-domain representation is one of the most fundamental operations in DSP.
- *Key Points:*
 - Use of libraries such as `numpy.fft` or `scipy.fft`.
 - Handling real vs. complex signals, windowing, zero-padding, etc.
 - Ensuring correct frequency bin mapping.

2. Inverse FFT (IFFT) of a Frequency Spectrum

- *Rationale:* Complements the FFT, allowing signals to be reconstructed in the time domain.
- *Key Points:*
 - Handling phase and amplitude.
 - Working with real vs. complex signals.
 - Managing array shapes, ensuring correct dimensionality.

3. Design and Application of a Digital Low-Pass Filter

- *Rationale:* Filters are a fundamental building block for noise reduction and band-limited analysis.
- *Key Points:*
 - IIR designs like Butterworth, Chebyshev, or Elliptic filters.
 - FIR design using windows (Hamming, Blackman, etc.).
 - Implementation details such as cut-off frequency, filter order, and passband ripple.

4. Design of a Bandpass Filter

- *Rationale:* Allows signals in a specified frequency band, rejecting others.
- *Key Points:*
 - Low and high cutoff frequencies.
 - Transition bands and stability considerations for IIR.
 - Zero-phase filtering vs. causal filtering (e.g. `filtfilt` vs. `lfilter`).

5. Resampling a Signal

- *Rationale:* Multi-rate DSP tasks (e.g., changing sampling rates for different system requirements).
- *Key Points:*
 - Upsampling vs. downsampling.
 - Polyphase filters or straightforward interpolation.
 - Handling boundary conditions to avoid aliasing.

6. Window-Based Smoothing or Moving Average

- *Rationale:* Reduces noise or fluctuations in the signal.

- *Key Points:*
 - Different window shapes (e.g. boxcar, triangular, Gaussian).
 - Managing boundary effects (padding vs. ignoring edges).
 - Implementation via convolution or direct sliding methods.

7. Computation of a Spectrogram

- *Rationale:* Essential time-frequency representation for non-stationary signals.
- *Key Points:*
 - Repeated short-time FFT with overlapping segments.
 - Windowing and step sizes (hop length).
 - Visualization in decibels or amplitude.

8. Peak Detection

- *Rationale:* Identifying local maxima for event detection or spectral peaks.
- *Key Points:*
 - Setting thresholds, minimum peak distances.
 - Handling noise and smoothing prior to detection.
 - Returning indices or amplitude values of identified peaks.

9. Cross-Correlation (Time or Frequency Domain)

- *Rationale:* Used for signal alignment, similarity, or system identification.
- *Key Points:*
 - Direct time-domain convolution vs. FFT-based multiplication.
 - Normalizing correlation or leaving it unnormalized.
 - Handling large data efficiently.

10. Wavelet Transform (e.g., DWT, CWT)

- *Rationale:* More advanced time-frequency analysis than STFT, excellent for non-stationary signals with transient features.
- *Key Points:*
 - Choice of wavelet families (Daubechies, Haar, Morlet, etc.).

- Discrete Wavelet Transform (DWT) vs. Continuous Wavelet Transform (CWT).
- Interpreting coefficients for denoising or feature extraction.

3 Conclusion

These ten tasks span a broad spectrum of typical signal-processing techniques, from fundamental frequency-domain transforms to advanced multi-rate and wavelet analyses. By testing large language models on each of these, one can gain insight into the model’s ability to:

- Generate correct, functional, and efficient code.
- Make appropriate use of well-known libraries (e.g., `numpy`, `scipy.signal`).
- Handle real-world edge cases in complex DSP scenarios.

A model that performs consistently well across these tasks demonstrates good potential for general-purpose coding in the signal analysis domain.