

[Accelerate](#) / Finding the component frequencies in a composite sine wave

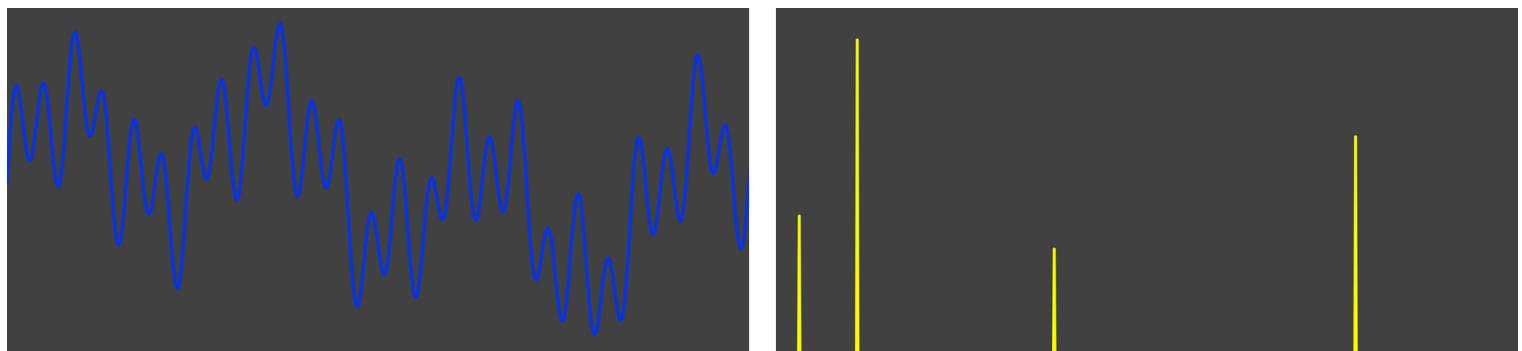
Article

# Finding the component frequencies in a composite sine wave

Use 1D fast Fourier transform to compute the frequency components of a signal.

## Overview

Accelerate's vDSP module provides functions to perform 1D fast Fourier transforms (FFTs) on vectors of data, such as audio signals. The example below shows an input signal (left) and its frequency domain representation (right) after transforming the signal with a forward FFT.



You can inspect the frequency-domain data of a forward FFT to compute the individual sine wave components of a composite wave. The technique described in this article is applicable to many digital signal processing applications, for example, finding the dominant frequencies in a dual-tone multi-frequency (DTMF) signal or removing noise from a signal.

## Synthesize a test signal

The function below generates a composite sine wave from a supplied array of component frequencies and amplitudes:

```

static func synthesizeSignal(frequencyAmplitudePairs: [(f: Float, a: Float)],  

                           count: Int) -> [Float] {  
  

    let tau: Float = .pi * 2  

    let signal: [Float] = (0 ..< count).map { index in  

        frequencyAmplitudePairs.reduce(0) { accumulator, frequenciesAmplitudePair in  

            let normalizedIndex = Float(index) / Float(count)  

            return accumulator + sin(normalizedIndex * frequenciesAmplitudePair.f *  

                tau)  

        }  

    }  
  

    return signal
}

```

## Create the composite signal

Create an array that contains frequency-amplitude tuples. You define the frequencies as the number of cycles per n. The highest measurable frequency, known as the Nyquist frequency, is the element with index  $n/2$ , which is 1023 in a zero-based array that contains 2048 elements.

The code below creates the array, `signal`, that contains four component sine waves:

```

let n = vDSP_Length(2048)  
  

let frequencyAmplitudePairs = [(f: Float(2), a: Float(0.8)),  

                               (f: Float(7), a: Float(1.2)),  

                               (f: Float(24), a: Float(0.7)),  

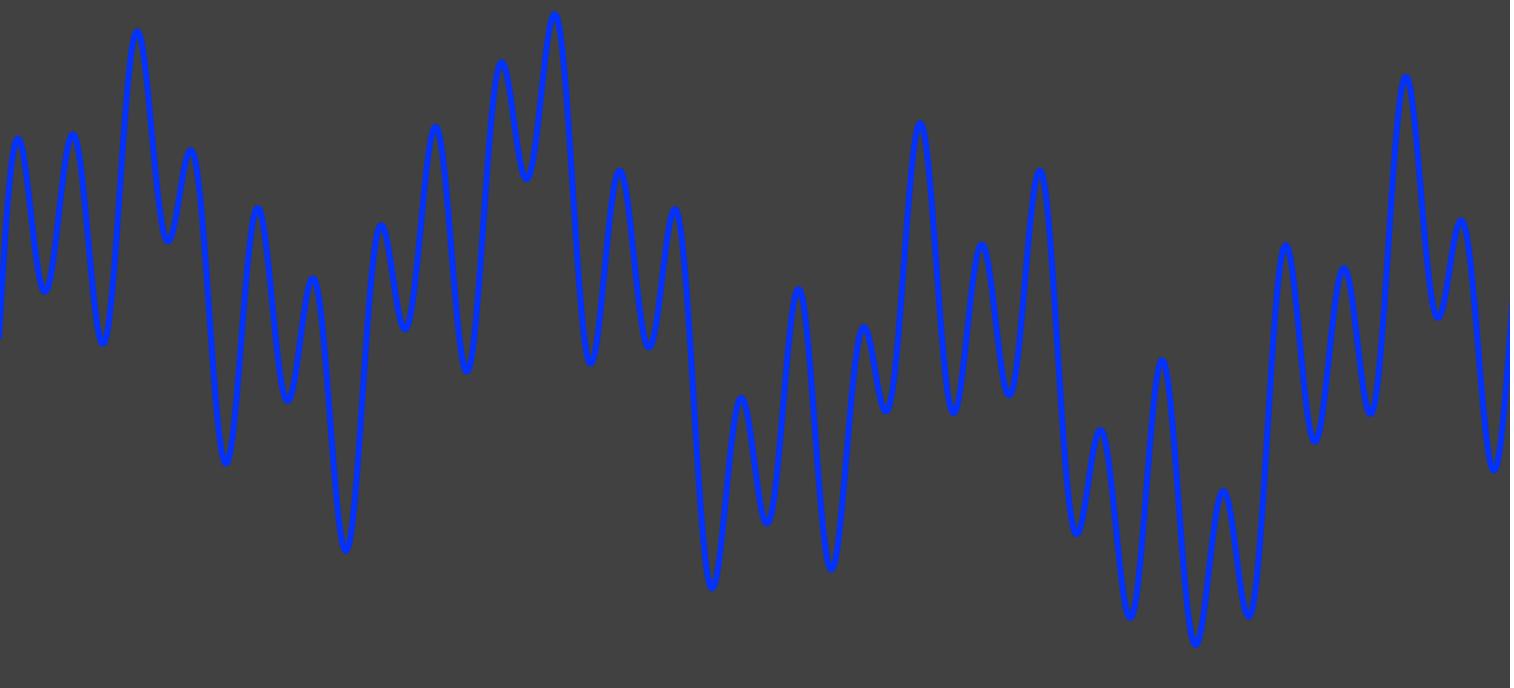
                               (f: Float(50), a: Float(1.0))]  
  

let signal = synthesizeSignal(frequencyAmplitudePairs: frequencyAmplitudePairs,  

                             count: Int(n))

```

The image below is a visualization of composite sine waves in `signal`:



## Create the FFT setup

Create a setup object that contains a precalculated weights array of complex exponentials required to perform the FFT operations. The values in the weights array simplify the FFT calculation. Creating this setup object can be expensive, so do it only once, for example, when starting your app. After creating the setup object, you can reuse it later.

The code below creates a setup object suitable for performing forward and inverse 1D FFTs on a signal containing  $n$  elements:

```
let log2n = vDSP_Length(log2(Float(n)))

guard let fftSetUp = vDSP.FFT(log2n: log2n,
                             radix: .radix2,
                             ofType: DSPSplitComplex.self) else {
    fatalError("Can't create FFT Setup.")
}
```

You can use this setup object for similarly sized smaller FFTs. However, using a weights array built for an FFT that processes a large number of elements can degrade performance for an FFT that processes a significantly smaller number of elements.

## Create the source and destination arrays for the forward FFT

The FFT operates on complex numbers. That is, it operates on numbers that contain a real part and an imaginary part. Create two arrays — one for the real parts and one for the imaginary parts — for the input and output to the FFT operation:

```
let halfN = Int(n / 2)

var forwardInputReal = [Float](repeating: 0,
                               count: halfN)
var forwardInputImag = [Float](repeating: 0,
                               count: halfN)
var forwardOutputReal = [Float](repeating: 0,
                               count: halfN)
var forwardOutputImag = [Float](repeating: 0,
                               count: halfN)
```

Because each complex value stores two real values, the length of each array is half that of signal.

## Perform the forward FFT

You use [DSPSplitComplex](#) structures to pass the separate real and imaginary arrays of the input and the output data to the FFT transform function.

The steps below perform the forward FFT:

1. Create a [DSPSplitComplex](#) structure to store a copy of signal that's represented as complex numbers.
2. Use [convert\(interleavedComplexVector:toSplitComplexVector:\)](#) to convert the real values in signal to complex numbers. The conversion stores the even values in signal as the real components in forwardInput, and the odd values in signal as the imaginary components in forwardInput.
3. Create a [DSPSplitComplex](#) structure with pointers to forwardOutputReal and forwardOutputImag to receive the FFT result.
4. Perform the forward FFT.

The code below shows how to perform the forward FFT using the steps described above:

```
forwardInputReal.withUnsafeMutableBufferPointer { forwardInputRealPtr in
    forwardInputImag.withUnsafeMutableBufferPointer { forwardInputImagPtr in
        forwardOutputReal.withUnsafeMutableBufferPointer { forwardOutputRealPtr in
            forwardOutputImag.withUnsafeMutableBufferPointer { forwardOutputImagPtr in
```

```

    // Create a `DSPSplitComplex` to contain the signal.
    var forwardInput = DSPSplitComplex(realp: forwardInputRealPtr.baseAddress,
                                       imagp: forwardInputImagPtr.baseAddress)

    // Convert the real values in `signal` to complex numbers.
    signal.withUnsafeBytes {
        vDSP.convert(interleavedComplexVector: [DSPComplex]($0.bindMemory),
                     toSplitComplexVector: &forwardInput)
    }

    // Create a `DSPSplitComplex` to receive the FFT result.
    var forwardOutput = DSPSplitComplex(realp: forwardOutputRealPtr.baseAddress,
                                         imagp: forwardOutputImagPtr.baseAddress)

    // Perform the forward FFT.
    fftSetUp.forward(input: forwardInput,
                     output: &forwardOutput)
}

}
}
}

```

On return, `forwardOutputReal` contains the real parts of the forward FFT, and `forwardOutputImag` contains the imaginary parts of the frequency-domain representation of the original signal.

## Compute component frequencies in the frequency-domain data

Use the `vDSP_zaspec` function to compute the autospectrum of the frequency-domain data in the `forwardOutputReal` and `forwardOutputImag` arrays. The autospectrum is the sum of squares of the complex and real parts of each complex frequency-domain element. The code below computes the autospectrum:

```

let autospectrum = [Float](unsafeUninitializedCapacity: halfN) {
    autospectrumBuffer, initializedCount in

    // The `vDSP_zaspec` function accumulates its output. Clear the
    // uninitialized `autospectrumBuffer` before computing the spectrum.
    vDSP.clear(&autospectrumBuffer)
}

```

```

forwardOutputReal.withUnsafeMutableBufferPointer { forwardOutputRealPtr in
    forwardOutputImag.withUnsafeMutableBufferPointer { forwardOutputImagPtr in
        var frequencyDomain = DSPSplitComplex(realp: forwardOutputRealPtr.baseAddress,
                                                imagp: forwardOutputImagPtr.baseAddress)
        vDSP_zaspec(&frequencyDomain,
                    autospectrumBuffer.baseAddress!,
                    vDSP_Length(halfN))
    }
}
initializedCount = halfN

```

The autospectrum of the forward FFT contains a series of high-magnitude items, rendered as vertical lines in the graph below:



The autospectrum values correspond to the frequencies and amplitudes you specified in the `frequencies` array. The code below scales the amplitudes to consider the autospectrum calculation and the inverse-transform step. To learn more about scaling time- and frequency-domain data, see [Understanding data packing for Fourier transforms](#).

```

let componentFrequencyAmplitudePairs = autospectrum.enumerated().filter {
    $0.element > 1
}.map {
    return ($0.offset, sqrt($0.element) / Float(n))
}

```

```

// Prints:
//      ["frequency: 2 | amplitude: 0.80", "frequency: 7 | amplitude: 1.20",
//      "frequency: 24 | amplitude: 0.70", "frequency: 50 | amplitude: 1.00"]

print(componentFrequencyAmplitudePairs.map {
    "frequency: \($0.0) | amplitude: \(String(format: "%.2f", $0.1))"
})

```

## Recreate the original signal

Use an inverse FFT to recreate a signal in the time domain, using the frequency-domain data returned by the forward FFT.

The steps below perform the inverse FFT:

1. Create the source of the inverse FFT, with pointers to `forwardOutputReal` and `forwardOutputImag`.
2. Create a `DSPSplitComplex` structure to receive the FFT result.
3. Perform the inverse FFT.
4. Return an array of real values from the FFT result. Because the forward transform has a scaling factor of 2 and the inverse transform has a scaling factor of the number of items, divide each result by `2 * n`:

```

var inverseOutputReal = [Float](repeating: 0,
                                count: halfN)
var inverseOutputImag = [Float](repeating: 0,
                                count: halfN)

let recreatedSignal: [Float] = forwardOutputReal.withUnsafeMutableBufferPointer { fo
    forwardOutputImag.withUnsafeMutableBufferPointer { forwardOutputImagPtr in
        inverseOutputReal.withUnsafeMutableBufferPointer { inverseOutputRealPtr in
            inverseOutputImag.withUnsafeMutableBufferPointer { inverseOutputImagPtr in

                // Create a `DSPSplitComplex` that contains the frequency-domain data
                let forwardOutput = DSPSplitComplex(realp: forwardOutputRealPtr.base,
                                                    imagp: forwardOutputImagPtr.base)

                // Create a `DSPSplitComplex` structure to receive the FFT result.
                var inverseOutput = DSPSplitComplex(realp: inverseOutputRealPtr.base,
                                                    imagp: inverseOutputImagPtr.base)
            }
        }
    }
}

```

```

    // Perform the inverse FFT.
    fftSetUp.inverse(input: forwardOutput,
                     output: &inverseOutput)

    // Return an array of real values from the FFT result.
    let scale = 1 / Float(n * 2)
    return [Float](fromSplitComplex: inverseOutput,
                  scale: scale,
                  count: Int(n))
}

}

}

}

```

On return, `recreatedSignal` is approximately equal to `signal`.

## See Also

### Fourier and Cosine Transforms

- 📄 Understanding data packing for Fourier transforms  
Format source data for the vDSP Fourier functions, and interpret the results.
- 📄 Performing Fourier transforms on interleaved-complex data  
Optimize discrete Fourier transform (DFT) performance with the vDSP interleaved DFT routines.
- 📄 Reducing spectral leakage with windowing  
Multiply signal data by window sequence values when performing transforms with noninteger period signals.
- {} Signal extraction from noise  
Use Accelerate's discrete cosine transform to remove noise from a signal.
- 📄 Performing Fourier Transforms on Multiple Signals  
Use Accelerate's multiple-signal fast Fourier transform (FFT) functions to transform multiple signals with a single function call.
- {} Halftone descreening with 2D fast Fourier transform  
Reduce or remove periodic artifacts from images.

### ☰ Fast Fourier transforms

Transform vectors and matrices of temporal and spatial domain complex values to the frequency domain, and vice versa.

### ☰ Discrete Fourier transforms

Transform vectors of temporal and spatial domain complex values to the frequency domain, and vice versa.

### ☰ Discrete Cosine transforms

Transform vectors of temporal and spatial domain real values to the frequency domain, and vice versa.